Nutrient stoichiometry shapes microbial coevolution

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ABSTRACT

Coevolution arises from the reciprocal genetic change between species and represents a major evolutionary force that contributes to the generation and maintenance of biodiversity. The tempo and mode of coevolution are affected by environmental conditions that modify species interactions. For example, the scarcity of essential elements influences the nutrition and productivity of host populations, which should not only regulate parasite dynamics, but also drive the evolution of defense and virulence traits. To explore the effects of nutrient availability on antagonistic coevolution, we conducted a long-term chemostat experiment where the marine cyanobacterium *Synechococcus* was challenged with a lytic phage under nitrogen (N) or phosphorus (P) limitation. Our manipulation of nutrient stoichiometry influenced the stability of host-parasite interactions, but also affected the underlying mode of coevolution. By assessing infectivity with more than 18,000 pairwise challenges using time-shift and network analyses, we documented directional selection for increased phage resistance, a pattern that is consistent with coevolutionary arms-race dynamics. In contrast, phage infectivity fluctuated through time, as expected when coevolution is determined by negative frequency-dependent selection, but was 70 % higher on naive hosts that evolved under N-limitation versus P-limitation. Furthermore, infection networks were 25 % more modular under P-limitation than N-limitation reflecting host-range contraction and an asymmetric coevolutionary trajectory. Together, our results demonstrate that nutrient stoichiometry creates eco-evolutionary feedbacks that may alter the dynamics and functioning of environmental and host-associated microbial communities.

SIGNIFICANCE STATEMENT

34  As obligate parasites, phage represent a significant source of mortality for marine

cyanobacteria, which are photosynthetic microorganisms that play a central role in the regulation

36 of biogeochemical processes, including energy flow, carbon sequestration, and the cycling of

nitrogen (N) and phosphorus (P). Phage also act as agents of evolutionary change by selecting

38 for resistant cyanobacteria, which can give rise to counter-resistant phage. This coevolutionary

process was differentially affected by the availability of N and P in ways that could impact the

40 ecology and evolution of one of the most abundant and functionally important groups of

microorganisms on Earth.

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46   In order to grow and reproduce, organisms must assimilate elements from the
environment to meet their nutritional and energetic demands. Ecological stoichiometry is a

48 theoretical framework that explicitly considers the mass balance of materials and energy in the
environment and the individuals that incorporate them (1). Of the approximately 25 elements

50 contained in biomass, nitrogen (N) and phosphorus (P) are two of the most limiting and essential
nutrients. N and P are needed for the synthesis of major macromolecules, including nucleic

52 acids, ribosomes, proteins, and cellular membranes that collectively influence an organism's
performance. However, the degree to which N and P are regulated greatly varies among major

54 groups of taxa. For example, the biomass stoichiometry of primary producers tends to be
flexible, reflecting the supply of nutrients in their environment (1). In contrast, the biomass

56 stoichiometry of consumer populations is homeostatically regulated and therefore tends to
remain constant even when the nutrient content of their resources fluctuates (1). Such differences

58 can create a nutritional imbalance between primary producers and consumers, which has
profound consequences for a wide range of ecological processes including resource competition,

60 host-parasite dynamics, and ecosystem functioning (2-4).

       Nutrient stoichiometry can also influence evolutionary processes. Nutrient limitation

62 often leads to a reduction in population size, which can diminish the efficiency of selection and
result in the accumulation of deleterious mutations through genetic drift (5). However, many

64 populations have adaptations that allow individuals to contend with absolute and relative
resource scarcity. For example, the disproportionate use of nucleotides that vary in N content

66 reduces stoichiometric mismatch between an organism and its environment (6). Similarly,
natural selection can operate on the material costs of gene expression in ways that lead to the

68   sparing of elements that are normally used in highly expressed proteins (7). These genetic

responses to nutrient limitation may give rise to the evolution of organismal stoichiometry. For

70   example, long-term carbon limitation led to increases in the N and P content of experimentally

evolved bacterial populations (8). However, evolutionary responses to nutrient stoichiometry

72   may depend on the identity of the limiting nutrient (9). While N limitation can select for stress-

related or catabolic genes with lower guanine-cytosine content (10), P limitation can favor the

74   replacement of phospholipids with sulfolipids in the cell membranes of certain microorganisms

(11).

76   In a community context, the combined effects of nutrient stoichiometry may give rise to

eco-evolutionary feedbacks. Such feedbacks occur when ecological interactions affect

78   evolutionary processes, which in turn modify species interactions and ecological dynamics (12).

For example, the rapid evolution of functional traits can produce diminished oscillations, longer

80   periods of cycling, and phase-shifted populations densities between hosts and their parasites, a

phenomenon referred to as "cryptic dynamics" (13). Theory suggests that cryptic dynamics can

82   arise when nutrient stoichiometry alters the stability of antagonistic species interactions (14),

which may ultimately intensify arms-race dynamics and negative frequency-dependent selection

84   (4). However, links among nutrient stoichiometry, eco-evolutionary feedbacks, and coevolution

remain to be tested.

86   Major advances in evolutionary ecology have been made through experimental studies of

microbial communities. In particular, bacteria and phage are ideal for studying eco-evolutionary

88   feedbacks owing to their large population sizes, rapid growth rates, and experimental tractability.

Moreover, bacteria and phage dynamics are critical for understanding the structure and function

90   of microbial food webs, especially in aquatic ecosystems (15). For example, *Synechococcus* is a

diverse and widely distributed group of primary producers in the world ocean with an estimated

92   global abundance of 10^{26} cells (16). While *Synechococcus* must contend with both N- and P-

limitation, it is also subject to a high degree of phage-induced mortality, which can lead to

94   coevolutionary dynamics (17). These coevolutionary processes can create feedbacks that not

only influence population dynamics, but also ecosystem functioning, including the turnover of N

96   and P (18).

In this study, we tested how nutrient stoichiometry affects host-parasite eco-evolutionary

98   dynamics in an experiment where a single genotype of marine *Synechococcus* was infected with

a phage strain in chemostats that were supplied with either N- or P-limited media. We isolated

100   hundreds of host and phage strains, which were challenged against one another to document

phenotypic changes in resistance and infectivity over the course of the experiment. Based on the

102   results from time-shift and network analyses, our data reveal that nutrient stoichiometry is a

bottom-up force that regulates eco-evolutionary feedbacks in ways that alter coevolutionary

104   processes.

RESULTS AND DISCUSSION

108   Our results demonstrate that the relative amounts of nitrogen (N) and phosphorus (P) in

an environment have strong eco-evolutionary effects on marine *Synechococcus* and its phage.

110   Communities were more stable under P-limitation than N-limitation, which likely was influenced

by the stoichiometric constraints and elemental homeostasis of the host and phage populations.

112   In addition to the importance of these ecological processes, resistant cyanobacteria rapidly

invaded the community, which was followed by the appearance of evolved phage with altered

114   host-ranges. The resulting patterns of resistance and infectivity were different among N- and P-

limited treatments suggesting that nutrient stoichiometry altered coevolutionary dynamics. Using

116 time-shift assays and network analyses, we detected strong signatures of arms-race dynamics in

both nutrient treatments. However, under P-limitation, we also observed patterns consistent with

118 negative frequency-dependent selection, which may reflect the underlying costs of maintaining

defense and virulence traits in different nutrient environments. Our findings generate testable

120 predictions regarding the coevolutionary mechanisms of local adaptation for microorganisms

found among oceans where N : P stoichiometry in known to systematically vary (19).

Eco-evolutionary effects of stoichiometry on community dynamics — Consistent with

124 predictions from the theory of ecological stoichiometry (1), we documented that *Synechococcus*

and phage dynamics were highly sensitive to variation in nutrient supply (RM-ANOVA; time x

126 stoichiometry, $P < 0.0001$). Under N-limitation, *Synechococcus* rapidly declined reaching its

minimum density 35 ± 11.2 days following phage addition ($1.7 \times 10^5 \pm 6.54 \times 10^4$ cells mL^{-1},

128 mean $\pm$ SEM, Fig. 1a, Table S4). This reduction in host abundance corresponded with a spike in

phage density, which peaked near 25 ± 8.0 days ($8.5 \times 10^8 \pm 1.40 \times 10^8$ particles mL^{-1}, mean $\pm$

130 SEM). Over the next 50 days, host densities slowly recovered, and entered a second round of

decline near day 100. In contrast, communities were more stable under P-limitation

132 (*Synechococcus*: t-test, $t_4 = -3.38$, $P = 0.04$; t-test, phage: $t_4 = 3.15$, $P < 0.001$, Fig. 1b, Table S4).

Hosts declined more slowly following phage addition (t-test, $t_4 = -4.00$, $P = 0.02$) and reached

134 their minimum density ($4.0 \times 10^5 \pm 5.74 \times 10^4$ cells mL^{-1}, mean $\pm$ SEM) near 78 ± 15.6 days.

Meanwhile, phage populations remained high and relatively constant ($2.4 \times 10^8 \pm 9.45 \times$

136 10^7 particles mL^{-1}, mean $\pm$ SEM) throughout the duration of the experiment. Nutrient

stoichiometry also affected the synchrony of the microbial populations within a chemostat.

138 While host and phage densities were out-of-phase with a lag of two to four days in N-limited

chemostats (r = -0.22 to -0.14, P < 0.001, Figs. 1, S1), there was no detectable coherence

between *Synechococcus* and phage experiencing P-limitation over the course of the experiment

(r = -0.04 to 0.013, P > 0.50, Figs. 1, S1).

Historically, ecologists have attempted to explain variation in population and community

dynamics assuming a single currency of resource. However, systems can exhibit a much wider

range of behaviors when resource stoichiometry is explicitly considered owing to feedbacks that

arise when interacting species vary in the degree to which their elemental composition is

homeoestatically regulated (20). For example, the molar N : P ratio of marine *Synechococcus*

converges upon the canonical Redfield value of 16 : 1 in some environments (21), but in other

instances can exceed 100 : 1 (22)reflecting the extremely plasticity of its biomass stoichiometry.

In contrast, phage are thought to have a fixed elemental composition owing to their relatively

simple structure, which minimally consists of genetic material (DNA or RNA) protected by a

proteinaceous capsid (23). Unless they contain auxiliary metabolic genes that allow for

redirection of resources (24), phage are entirely dependent on the stoichiometry of their host

when assembling new viruses. Biophysical models predict that the T4-like phage used in our

study have a high P demand due to their low N : P ratio of approximately 7 : 1 (23). As a

consequence, phage productivity is thought to be constrained by host P content. For example,

host lysis was delayed by 18 hrs leading to an 80 % reduction in phage burst size when P-limited

Synechococcus WH7803 was infected with a myorvirus (25). Such findings suggest that nutrient

stoichiometry can profoundly shape host-phage dynamics and provide an ecological explanation

for the contrasting dynamics observed in our study (Fig. 1).

In addition, nutrient stoichiometry may affect community dynamics by generating eco-

evolutionary feedbacks. Such feedbacks arise when there are rapid changes in host or parasite

162   traits leading to cryptic dynamics, which are not predicted from traditional ecological theory

(13). Cryptic dynamics require the coexistence of multiple host genotypes that exhibit trade-offs

164   in competitive ability and parasite defense (13). While there are many intracellular and

extracellular mechanisms that bacteria can employ to resist phage infection, *Synechococcus*

166   appears to undergo mutations of receptors on the cell surface that reduce or eliminate attachment,

thus precluding entry of the virus into the host (26). Resistance mutations in *Synechococcus* are

168   often accompanied by a reduction in growth rate, the magnitude of which can vary depending on

the identity of the host and virus (27). These fitness costs have important consequences for

170   understanding community stability. Without a reduction in growth rate, resistant *Synechococcus*

would outcompete the sensitive host and drive the phage population extinct. Moreover, fitness

172   costs establish a trade-off that satisfies the requirement for cryptic dynamics to emerge (13). In

sum, our data support the view that nutrient stoichiometry altered host-phage dynamics and

174   stability via eco-evolutionary feedbacks.

176   **Nutrient-dependent coevolution —** Nutrient stoichiometry had strong, but asymmetric effects

on microbial coevolution. Phage-resistant cyanobacteria rapidly arose and swept through the

178   population over the course of the experiment (Fig. S2). Within nine days of phage addition, ~30

% of the *Synechococcus* strains were resistant to the ancestral phage in both the N- and P-limited

180   chemostats. Although average resistance continued to increase over time (RM-ANOVA, $F_{6, 97} =$

14.27, $P < 0.001$), it was not affected by nutrient stoichiometry ($F_{6, 97} = 0.51$, $P = 0.30$). In

182   contrast, changes in phage infectivity were significantly altered by nutrient stoichiometry. For

example, we recovered a host-range mutant from a P-limited chemostat (day 129) that was able

184   to infect a *Synechococcus* strain (day 166), which was resistant to the ancestral phage. In

addition, three phage strains from N-limited chemostats and two phage strains from P-limited

186   chemostats (day 166) were able to infect phage-resistant *Synechococcus* isolated from other

chemostats earlier in the study. Overall, phage from the N-limited chemostats were 12 % more

188   infective than phage from the P-limited chemostats (RM-ANOVA, stoichiometry x time: $F_{4,\ 166}$ =

4.83*, P* = 0.001). These effects of nutrient stoichiometry on host-phage interactions led us to

190   further explore the potential mechanism and patterns underlying the observed coevolution in our

system.

192   There are two primary modes by which antagonistic coevolution is thought to occur. The

first is through arms-race dynamics involving gene-for-gene specificity, where directional

194   selection leads to an escalation of resistance and infectivity. Arms-race dynamics were originally

described for coevolving populations of plants and pathogens, but since then have been

196   commonly reported in studies of bacteria and phage (28, 29). The second mode of antagonistic

co-evolution involves negative frequency-dependent selection where parasites evolve to infect

198   common hosts, which in turn favors rare host alleles. Negative frequency-dependent selection is

often associated with infections that require matching alleles and is well documented in

200   invertebrate systems (30), but has also been described in some studies of bacteria and phage(31).

While arms-race dynamics and negative frequency-dependent selection are often viewed as

202   occupying different ends of the coevolutionary spectrum, they are not mutually exclusive (32). In

fact, evidence suggests that over time, arms-race dynamics can give way to negative frequency-

204   dependent selection, in some cases depending on resource availability (33).

        One powerful way to discern modes of coevolution is through the use of time-shift

206   analyses. This approach involves determining the success of infections for combinations of hosts

and parasites that are isolated from different time points in a controlled experiment or other

208   longitudinal type of study (34). When applied to our chemostat data, time-shift analyses

indicated that coevolutionary dynamics were significantly affected by nutrient stoichiometry

210   (Fig. 3, RM-ANOVA, stoichiometry x time, $F_{1, 601} = 4.26$, $P = 0.039$). Specifically, when hosts

isolated from the phage-amended chemostats were challenged against current and past phage,

212   infectivity was weak owing to the evolution of phage resistance (Fig. 3 a, b). Infectivity was

stronger when hosts were challenged against future phage, but only for *Synechococcus* strains

214   that were isolated earlier in the experiment (days -6, 9, and 23). Consistent with arms-race

dynamics, these patterns reflect escalating resistance in the host population that were not affected

216   nutrient stoichiometry (Fig. 3 a, b; stoichiometry x time shift, $F_{1, 602} = 0.05$, $P = 0.82$). We

obtained additional insight by conducting time-shift analyses with *Synechococcus* that were

218   isolated from no-phage control chemostats (Fig. 3 c, d; stoichiometry x time shift, $F_{1, 289} = 10.3$,

$P = 0.0015$). In contrast to initial expectations, phage infectivity was not uniformly high on these

220   naive hosts, but rather fluctuated over time (Fig. 3 c, d) owing in part to host-range contraction.

We also found that infectivity was 70 % stronger under N-limitation (0.61 ± 0.289) than P-

222   limitation (0.36 ± 0.332), which together is consistent with nutrient-dependent coevolution that

was driven by negative frequency-dependent selection.

224   Network analyses provide an additional way to evaluate the effects of nutrient

stoichiometry on the mode of coevolution. With this approach, one can estimate the degree of

226   nestedness that exists among pairs of hosts and phage relative to randomized data. Often,

infection data from bacteria-phage systems are highly nested (35), which means they contain

228   ordered subsets of phenotypes whereby hosts from later time points are resistant to earlier

phages, and phage from later time points are able to infect earlier hosts (36). In our study,

230   infection networks were significantly nested, a pattern that arises from arms-race dynamics.

However, the degree of nestedness was not affected by nutrient stoichiometry (NODF, $t_4 = 0.37$,

232     $P = 0.72$, Table S5). Infection networks can also exhibit modularity, which is more consistent

with negative frequency-dependent selection(36). Modules reflect dense clusters of interacting

234   host and phage compared to other strain combinations found in a bipartite infection matrix.

Although less commonly documented, modularity has been observed in large-scale oceanic

236   surveys of bacteria and phage, but has been attributed local adaptation and the phylogenetic

breadth of the hosts (37). However, theory suggests that modularity can emerge *via* coevolution

238   between two populations at the local scale. Using a "relaxed lock and key" model,

simultaneously nested and modular structures arose under simple chemostat conditions that

240   assumed gene matching between phage tail-fibers and host receptors (38). Our results are in

agreement with these predictions, but suggest that network properties are affected by host

242   nutrition. Specifically, we found that host-phage interactions were 50 % more modular under P-

limitation than N-limitation ($t_{3.0018} = -3.55$, $P = 0.038$, Fig. 4, Table S5,) suggesting that nutrient

244   stoichiometry constrains host-phage interactions leading to increased specialization. Such

findings are consistent with the view that resources can influence the modes of coevolution based

246   in part on the fitness costs associated with host defense and infection strategies (33)

       Mechanistically, there are many ways that nutrient stoichiometry could shape bacteria-

248   phage coevolution. For example, the "dangerous nutrients" hypothesis predicts that trajectories

of coevolution can be influenced when receptors used by nutrient-limited hosts also serve as the

250   targets of phage adsorption (39). However, it does not appear that myoviruses, including the

strain used in this study, attach to the protein receptors of cyanobacteria that are used for nutrient

252   transport. Instead, evidence from whole-genome sequencing suggests that phage-resistant

Synechococcus accumulate mutations in hypervariable genomic islands that encode for

254   lipopolysaccharide (LPS) (17), a major component of the outer membrane in Gram-negative

bacteria. The structural complexity of LPS is affected by nutrient limitation (40) and such

256   changes in molecular structure of the cell membrane can interfere with phage adsorption leading

to resistance (41). Nevertheless, modification of tail-fibers allow phage to overcome resistance of

258   marine cyanobacteria in some instances (42) while the acquisition of host-like P-assimilation

genes may aid in the successful infection of nutrient-limited hosts (43).

Conclusions — Most organisms live in environments where they are limited by the relative or

262   absolute amount of one or more essential resources. It is well established that such variation in

nutrient stoichiometry can regulate ecological phenomena ranging from species interactions to

264   ecosystem-level processes. We demonstrated that nutrient stoichiometry also affects the eco-

evolutionary dynamics of microbial communities through directional selection for increased

266   phage resistance and the evolution of host-range mutations. Identifying the targets of selection in

contrasting nutrient environments will help elucidate the genetic mechanisms of coevolution and

268   the trade-offs associated with defense and virulence traits. Such findings will provide valuable

insight into the evolutionary ecology of marine food webs. While our study offers promising

270   avenues to better understand bacteria-phage interactions in the oceans, nutrient stoichiometry is

also important in determining the nutrition, health, and disease susceptibility for for non-

272   microbial hosts (2). By considering the ecology and evolutionary effects of nutrient

stoichiometry, we may be able to better manage the persistence, emergence, and evolution of

274   infectious diseases.

276   **MATERIALS AND METHODS**

Strains and media — We evaluated the effects of nutrient stoichiometry on the eco-

278   evolutionary dynamics of the marine cyanobacterium *Synechococcus* WH7803 and a lytic T4-

like phage belonging to the Myoviridae family of phage (SRIM8). Changes to nutrient supply

280   can alter the equilibrium density of microbial populations in continuous culture (chemostats),

which in turn, may affect cyanobacteria-phage contact rates (45). Therefore, we adjusted both the

282   concentrations and ratios of N and P in modified AN artificial seawater medium (Tables S1-S3)

to induce N- or P-limitation while maintaining similar equilibrium densities of *Synechococcus*

284   between the stoichiometry treatments prior to phage addition. Specifically, the N-limited

medium had an N : P ratio of 10 : 1 (KNO_3; N = 220 µM) while the P-limited medium had a N :

286   P ratio of 40 : 1 (K_2HPO_4; P = 11 µM). Growth assays confirmed that *Synechococcus* was

limited by N under low N : P supply and by P under high N : P supply (see Fig. S3).

Chemostat experiment — We supplied ten chemostats, each with a 40 mL operating volume,

290   with N-limited or P-limited medium at a dilution rate of 1 d^{-1} (Tables S1-S3). The chemostats

were maintained in a Percival growth chamber (Perry, IA, USA) at 25 °C on a 14:10 light : dark

292   cycle under 20 µE m^{-2} s^{-1} and homogenized with magnetic stir bars following inoculation with a

single-colony strain of *Synechococcus* WH7803. We allowed the *Synechococcus* populations to

294   equilibrate in the chemostats for 125 d prior to initiating the "phage-amended" treatment by

introducing an aliquot of a plaque-purified S-RIM8 to three randomly chosen chemostats in each

296   nutrient treatment to achieve a multiplicity of infection (phage : cyanobacteria ratio) of

approximately 10. To document the potential influence of stoichiometry on population dynamics

298   and population size, we maintained two "no phage" chemostats with only *Synechococcus* in both

stoichiometry treatments (Fig. S4).

Community dynamics — We tracked *Synechococcus* and phage densities in each chemostat via epifluorescent microscopy every other day for 172 days. *Synechococcus* populations were enumerated by concentrating samples from a chemostat onto 0.22-μm nominal pore-size black polycarbonate filters. After removing cellular material (0.2 μm filtration) and extracellular DNA with DNase I, we concentrated samples onto 0.02 μm Anodisc filters for phage enumeration. Each phage-containing filter was then stained with SYBR Green I. We captured and analyzed ten images from each filter with a CY3 filter set (ex: 550 nm, em: 570 nm) for *Synechococcus* or a FITC filter set (ex: 497 nm, em: 520 nm) for phage using a Zeiss microscope and Axiovision imaging software (Release 4.5 SP1).

We evaluated the effects of nutrient stoichiometry on *Synechococcus* and phage dynamics in three ways. First, we tested for the main and interactive effects of nutrient stoichiometry using repeated measures (RM) ANOVA. Second, because cryptic dynamics can alter the degree of synchrony between hosts and parasites, we performed cross-correlation analyses on pre-whitened population data using Auto Regressive Integrated Moving Average (ARIMA) procedures in SAS (18). Last, we estimated the effects of nutrient stoichiometry on the stability of *Synechococcus* and phage populations as the inverse of the coefficient of variation (CV) over time using *t*-tests.

Co-evolutionary dynamics — To test for the effects of nutrient stoichiometry on phenotypic coevolution, we tracked changes in infection patterns between *Synechococcus* and its phage over time. We isolated multiple (3-5) *Synechococcus* strains from each chemostat using dilution-to-extinction techniques six days prior to phage addition (day -6) and at days 9, 23, 72, 129, 148,

and 166 after phage addition. Single-colony strains of *Synechococcus* were harvested near mid-

324   log phase then concentrated *via* centrifugation, preserved in glycerol (10 % final concentration),

and stored at -80 °C until reanimation. We also isolated multiple (3-5) phage strains from the

326   phage-amended chemostats on days 23, 72, 129, 148, and 166 through plaque purification using

ancestral *Synechococcus* WH7803 as the host. Each 1 µm syringe-filtered phage lysate was

328   preserved in glycerol (10 % final concentration) at -80 °C.

With the isolated chemostat strains, we quantified host resistance and phage infectivity

330   using challenge assays. Each pair-wise challenge was completed in triplicate by adding 20 µL of

a phage stock (~10^7 particles mL^{-1}) to 200 µL of a dilute *Synechococcus* strain (~10^6 cells mL^{-1})

332   in 96-well plates. Challenge assays were performed using the same medium from which the host

strain was originally isolated. Turbid cultures of *Synechococcus* WH7803 appear bright pink

334   owing to the intracellular photosynthetic pigment phycoerythrin. When infected by SRIM8,

however, cells are lysed and cultures become clear (18). Based on this, we scored each

336   *Synechococcus* strain as resistant (and the phage strain as infective) if there was a lack of growth

after a two-week incubation under continuous light (20 µE m^{-2} s^{-1}) at 25 °C compared to control

338   wells (n = 3) that contained heat-killed phage. In total, there were 18,050 pairwise challenges

between chemostat-isolated strains of *Synechococcus* and phage that resulted in a bipartite

340   infection matrix, which we used for all analyses related to phenotypic coevolution.

To quantify the effect of stoichiometry on coevolutionary dynamics, first, we used RM-

342   ANOVA to test for trends in average resistance and average infectivity over the course of the

chemostat experiment. Second, we used the infection matrix to perform time-shift analyses,

344   which involved calculating the proportion of successful infections that occurred between hosts

that were challenged against past, contemporary, and future phage strains (34). We statistically

346 analyzed the time-shift data using RM-ANOVA with nutrient treatment, phage treatment, and

time as fixed effects, while the chemostat replicate identifier was treated as a random effect with

348 a corARMA covariance matrix (46). To visualize the data, contemporary challenges were

centered at a time-shift of zero, while interactions with past phage were represented in negative

350 space and future interactions were represented in positive space. Last, to gain insight into

potential mechanisms underlying stoichiometrically driven coevolution, we used community

352 network analyses to calculate the connectance, nestedness, and modularity for each chemostat

infection matrix. Connectance was calculated as the number of interactions divided by network

354 size. We calculated nestedness using the NODF metric, which ranges from 0 (non-nested) to 1

(perfectly nested) and normalizes for matrix size. We used the LP-BRIM algorithm to find the

356 partition that maximizes Barber's modularity (Qb), which ranges from 0 (all interactions are

between modules) to 1 (all interactions are within modules). The network statistics were

358 calculated using 100,000 random Bernoulli simulations in the BiWeb package for MATLAB (35,

47, http://github.com/tpoisot/BiWeb). We then tested for differences in connectance, nestedness,

360 and modularity between the nutrient treatments using t-tests.

362 **Available code and data.** Data and code are available in the public GitHub repository

https://github.com/LennonLab/eco-evo-stoich

Author contributions

366 MLL, SWW, and JTL designed study; MLL performed research; MLL and JTL analyzed data;

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 FIGURE CAPTIONS

Fig. 1. Microbial community dynamics were affected by nutrient stoichiometry. *Synechococcus*

492   and phage densities were tracked in replicate (n = 3) chemostats receiving N- or P-limited media.

Vertical lines at day 0 indicate time of phage amendment. See Fig. S4 for *Synechococcus*

494   dynamics in the no-phage control chemostats. Data are represented as mean ± SEM.

496   **Fig. 2** Phenotypic co-evolution between hosts (*Synechococcus*) and phage was affected by

nutrient stoichiometry. We calculated infectivity based on the proportion of successful infections

498   between *Synechococcus* strains and phage strains that were isolated from chemostats at different

time points. Infectivity is proportional to the width of the edges (lines) connecting nodes

500   (symbols). Black squares correspond to phage isolated from the phage-amended chemostats,

white circles correspond to *Synechococcu*s isolated from phage-amended chemostats, and grey

502   circles correspond to naive *Synechococcus* isolated from no-phage control chemostats. The

absence of a line indicates that *Synechococcus* isolates were resistant to a particular phage

504   challenge.

506   **Fig. 3.** Time-shift analysis of host-phage infectivity reveals the effects of stoichiometry on

coevolution. Contemporary interactions (i.e., those between a host and phage strain isolated at

508   the same time point) are centered at time zero (grey vertical line) along the time-shift

(horizontal) axis. Interactions with past phage are shifted to the left (negative values) and

510   interactions with future phage are shifted to the right (positive values). Each black line

corresponds with the mean infectivity for *Synechococcus* isolated from a specific time point as

512   indicated by the open circle containing the isolation day (-6, 9, 23, 72, 100, 129, or 166). When

comparing challenges between hosts and phage from phage-amended chemostats (a, b),

514   infectivity was weak for hosts that were isolated after day 23 or when challenged against phage

from the past owing to the evolution of resistance. Such findings are consistent with arms-race

516   dynamics where directional selection gives rise to escalating host resistance. We also challenged

phage against naive hosts from the no-phage control chemostats (c, d). From this, we found that

518   infectivity was significantly higher under N-limitation than P-limitation, but overall, was lower

than what would be expected under arms-race dynamics. Instead, the fluctuations in infectivity

520   with respect to time-shift are consistent with negative frequency-dependent selection and reflect

asymmetry in the coevolution between *Synechococcus* and phage.

Fig. 4. Host-phage infection networks based on interactions between *Synechococcus* and phage

524   isolates that coevolved under N- and P-limitation. Networks were significantly nested, consistent

with expectations of host-phage systems coevolving under arms-race dynamics. However, the

526   degree of nestedness was not affect by stoichiometry. Left panel: Networks were also

significantly modular when compared to randomized networks, but the degree of modularity was

528   significantly greater in P-limited networks. Right panel: Infection networks with interactions

(colored cells) rearranged using the LPBrim algorithm in BiWeb to reflect the modular structure.

530   Each colored grouping within a panel corresponds to a calculated module within the interaction

network of N-limited (a – c) or P-limited (d – f) chemostats.

534   **Fig. 1**

Fig. 1. Microbial community dynamics were affected by nutrient stoichiometry. Host (*Synechococcus*) and phage densities were tracked in replicate (n = 3) chemostats receiving N- or P-limited media. Vertical lines at day 0 indicate time of phage amendment. See Fig. S4 for *Synechococcus* dynamics in the no-phage control chemostats. Data are represented as mean ± SEM.

Fig. 2

Figure 2 Phenotypic co-evolution based on infectivity of phage strains on *Synechococcus* strains isolated from N- and P-limited chemostats. We calculated infectivity based on the proportion of successful infections between hosts and phage that were isolated from chemostats at different time points. Infectivity is proportional to the width of the edges (lines) connecting nodes (symbols). Black squares correspond to phage isolated from the phage-amended chemostats, white circles correspond to *Synechococcus* isolated from phage-amended chemostats, and grey circles correspond to naive *Synechococcus* isolated from no-phage control chemostats. The absence of a line indicates that *Synechococcus* isolates were resistant to phage challenge.

Fig. 3. Time-shift analysis of host-phage infectivity reveals the effects of stoichiometry on coevolution. Contemporary interactions (i.e., those between a host and phage strain isolated at the same time point) are centered at time zero (grey vertical line) along the time-shift (horizontal) axis. Interactions with past phage are shifted to the left (negative values) and interactions with future phage are shifted to the right (positive values). Each black line corresponds with the mean infectivity for *Synechococcus* isolated from a specific time point as indicated by the open circle containing the isolation day (-6, 9, 23, 72, 100, 129, or 166). When comparing challenges between hosts and phage from phage-amended chemostats (a, b), infectivity was weak for hosts that were isolated after day 23 or when challenged against phage from the past owing to the evolution of resistance. Such findings are consistent with arms-race dynamics where directional selection gives rise to escalating host resistance. We also challenged phage against naive hosts from the no-phage control chemostats (c, d). From this, we found that infectivity was significantly higher under N-limitation than P-limitation, but overall, was lower than what would be expected under arms race dynamics. Instead, the fluctuations in infectivity with respect to time-shift are consistent with negative frequency-dependent selection and reflect asymmetry in the coevolution between *Synechococcus* and phage.

Fig. 4

Fig. 4. Host-phage infection networks based on interactions between *Synechococcus* and phage isolates that coevolved under N- and P-limitation. Networks were significantly nested, consistent with expectations of host-phage systems coevolving under arms race dynamics. However, the degree of nestedness was not affect by stoichiometry. Left panel: Networks were also significantly modular when compared to randomized networks, but the degree of modularity was significantly greater in P-limited networks. Right panel: Infection networks with interactions (colored cells) rearranged using the LPBrim algorithm in BiWeb to reflect the modular structure. Each colored grouping within a panel corresponds to a calculated module within the interaction network of N-limited (a – c) or P-limited (d – f) chemostats.