Cumulative effects of mutation and selection on susceptibility to bacterial pathogens in

Caenorhabditis elegans

Sayran Saber[1,2,*], Lindsay M. Johnson[1,3,*], Md. Monjurul Islam Rifat[1], Sydney Rouse[1], and

Charles F. Baer[1,4,†]

1 – Dept. of Biology, University of Florida, Gainesville, FL 32611 USA

2 – Present address: Dept. of Biological Science, Florida International University, Miami, FL

33199 USA

3 – Present address: Dyne Therapeutics, Waltham, MA 02451 USA

4 - University of Florida Genetics Institute, Gainesville, FL 32611 USA

* - These authors contributed equally

[†] - Communication to: cbaer@ufl.edu

Abstract

Understanding the evolutionary and genetic underpinnings of susceptibility to pathogens is of fundamental importance across a wide swathe of biology. Much theoretical and empirical effort has focused on genetic variants of large effect, but pathogen susceptibility often appears to be a polygenic complex trait. Here we investigate the quantitative genetics of survival over 120 hours of exposure ("susceptibility") of *C. elegans* to three bacterial pathogens of varying virulence, along with the standard laboratory food, the OP50 strain of *E. coli*. We compare the genetic (co)variance input by spontaneous mutations accumulated under minimal selection to the standing genetic (co)variance in a set of 47 wild isolates. Three conclusions emerge. First, mutations increase susceptibility to pathogens, and susceptibility is uncorrelated with fitness in the absence of pathogens. Second, the orientation in trait space of the heritable (co)variance of wild isolates is sufficiently explained by mutation. However, with the possible exception of *S. aureus*, pathogen susceptibility is clearly under purifying, directional, selection of magnitude roughly similar to that of competitive fitness in the MA conditions. The results provide no evidence for fitness tradeoffs between pathogen susceptibility and fitness in the absence of pathogens.

Introduction

The relationship between pathogens and their hosts holds an exalted place in evolutionary

biology (Haldane 1949; Hamilton 1980). It is well-established that susceptibility of individuals to

the harmful effects of pathogens often has a genetic basis. The notion that *particular* multilocus

host genotypes confer resistance or susceptibility to *particular* pathogen genotypes has been

especially important, because "Red Queen" models of the evolution of sex and recombination

are predicated on fluctuating epistasis (Barton 1995), with negative frequency-dependent

selection resulting from a host-pathogen arms-race providing the most obvious plausible

scenario (Hamilton 1980).

In some cases, the genetic basis of variation in host susceptibility to pathogens can be

attributed to variants at one or two loci. For example, variants at a few loci in the human

genome (e.g., *HBB*, *ABO*, *G6PD*) are famously implicated in resistance to malaria (Rockett et

al. 2014). Often, however, the heritable basis of susceptibility to a particular pathogen appears

to be polygenic, with a substantial fraction of unexplained heritability (Lefebvre and Palloix 1996;

Hill 2001; Daub et al. 2013; Cogni et al. 2016), sometimes even in cases with segregating large-

effect loci (e.g., malaria; Band et al. 2019). Pathogens may thus impose significant selection

even in the absence of Red Queen dynamics. It seems reasonable, then, to consider

"susceptibility to pathogen X" as a generic complex trait; in some cases, there will be loci of

large effect that explain much of the variance in susceptibility, but other times there will not.

Two fundamental issues in understanding the biology of complex traits, are (*i*) what is

the rate of input of genetic variance (and covariance) by mutation? and (*ii*) what is the

relationship of the trait(s) with fitness? Taken together, these two considerations present a

hierarchy of evolutionary hypotheses. The simplest explanation for any pattern of (co)variation

is neutral evolution *sensu stricto*, i.e., the pattern can be sufficiently explained by neutral

mutation and random genetic drift. If the pattern cannot be plausibly explained by neutral

evolution, the next-simplest explanation is purifying selection against unconditionally deleterious

mutations, plus drift. Only if those two explanations prove unsatisfactory need more

complicated explanations be considered; for example, scenarios of balancing or positive

selection.

For quantitative traits, the analog of the per-nucleotide mutation rate, μ, is the mutational

variance $V_M = U\alpha^2$, where $U = \sum_i \mu_i$ is the genomic mutation rate and α^2 is the average squared

effect of a new mutation on the trait (Clayton and Robertson 1955; Lynch and Hill 1986; Barton

1990). For a neutral trait at mutation-drift equilibrium, the genetic variation segregating within a

population $V_G = 2N_e V_M$, and populations will diverge asymptotically at rate $2V_M$ (Lynch and Hill

1986). For traits under purifying selection (directional or stabilizing), at mutation-selection

balance (MSB), $V_G \approx \frac{V_M}{s}$, where s is the average strength of purifying selection against a mutant

allele that affects the trait (Barton 1990; Kondrashov and Turelli 1992; Charlesworth 2015). The

relationship $V_G \approx \frac{V_M}{s}$ implicitly includes both the direct selective effects of mutations on the trait

in question, but also the pleiotropic effects of selection on correlated traits.

The multivariate extensions of V_G and V_M are the standing genetic and mutational

variance-covariance matrices, **G** and **M**, of which the diagonal elements are the variances, $V_{G,i}$

and $V_{M,i}$ respectively, and the off-diagonal elements are the genetic/mutational covariance

between traits i and j, $Cov_{G,ij}$ and $Cov_{M,ij}$ (Lande 1980; Phillips and McGuigan 2006). The

covariances capture the cumulative effects of pleiotropy and linkage disequilibrium.

Here we report results of a set of experiments in which we (*i*) estimate the cumulative

effects of spontaneous mutations on the susceptibility of two strains of *C. elegans* to three

bacterial pathogens, and (*ii*) estimate the standing genetic (co)variance for susceptibility to

those same three pathogens from a set of ~50 *C. elegans* wild isolates. Our first, broad goal is

to determine the realms of consistency and of idiosyncrasy in the mutational process in the

context of host-pathogen evolution – what features are specific to particular host genotypes, or

particular pathogens, and what features are general. Our second, broad goal is to draw

inferences about the strength and pattern of natural selection acting on *C. elegans*' susceptibility

to bacterial pathogens, relative to the inferred strength of selection on competitive fitness.

Materials and Methods

Mutation Accumulation (MA) experiments – basic principles

An MA experiment is simply a pedigree, in which replicate populations derived from a known,

(ideally) isogenic progenitor are propagated under conditions in which natural selection is

minimized (Mukai 1964; Halligan and Keightley 2009). Typically, selection is minimized by

minimizing N_e; mutations with fitness effects $s<1/4N_e$ are effectively neutral (Keightley and

Caballero 1997). MA experiments can be used to illuminate the workings of natural selection in

two ways. First, they provide the most straightforward and least assumption-laden way to

estimate mutational (co)variances, which can be compared to the standing genetic

(co)variances as outlined in the Introduction (**Supplemental Figure S1**). Second, if selection is

directional, the change in the trait mean (the mutational bias, ΔM) will point in the opposite

direction of selection. On average, fitness declines under MA conditions. Traits that are

positively correlated with fitness should, on average, exhibit a negative mutational bias (ΔM<0);

traits that are negatively correlated with fitness should, on average, exhibit a positive mutational

bias (ΔM>0).

Mutation Accumulation (MA) lines

Details of the mutation accumulation experiment are given in Baer et al. (2005). N2 is the

standard laboratory strain of *C. elegans*; PB306 is a wild isolate graciously provided by Scott

Baird. The basic protocol follows that of Vassilieva and Lynch (1999) and is depicted in

Supplemental Figure S2. Briefly, replicate populations (MA lines) were initiated from a

cryopreserved stock of a highly-inbred ancestor and propagated by transferring a single

immature hermaphrodite at each generation (four-day intervals) under standard laboratory

conditions. The lines were propagated for 250 transfers (G_{250}) prior to cryopreservation. Data

on transfers and additional context is given in **Supplemental Table S1**. Of the 100 N2 and

PB306 MA lines initiated in 2001, we report results from 59 N2 and 65 PB306 lines. On

average, each MA line carries about 65 spontaneous base-substitution and small indel

mutations (Rajaei et al. 2021), with a few larger structural variants (Saxena and Baer,

unpublished results). The average number of mutations per genome is nearly identical between

the N2 and PB306 MA lines, thus any differences between the two strains in the cumulative

effects of mutations are the result of differences in the distribution of mutational (and/or

epimutational) effects, rather than in the total number of mutations.

The cryopreserved progenitor (G0) serves as a control in two ways. The difference

between the trait mean of the G0 and MA treatments (ΔM) quantifies the cumulative effect of

spontaneous mutations on the trait mean. If the G0 progenitor is subdivided into replicate

"pseudolines" (PS) which are subsequently treated identically to the MA lines (**Supplemental**

Figure S2), the difference between the among-MA line and the among-PS line components of

variance represents the cumulative heritable variance resulting from the accumulation of

spontaneous mutations (and potentially epimutations, depending on the experimental design;

Beltran et al. 2020; Johnson et al. 2020). PS lines were constructed by thawing a sample of the

cryopreserved G0 progenitor and picking individual L4 stage hermaphrodites to new plates,

each of which was designated a PS line. PS lines were subsequently treated as if they were MA

lines. Unfortunately, due to a management error on the part of the senior author (CFB), not all

pathogen treatments include similar numbers of pseudolines (see below).

Competitive fitness assay

Competitive fitness of the MA lines and their G0 progenitors was assayed in May, 2005. The

details of that assay have been published elsewhere (Yeh et al. 2018), and are reprised in

section 4 of the **Supplemental Material**. Fitness data are given in **Supplemental Table S2**.

Wild isolates

A collection of wild isolates of *C. elegans* was obtained from Erik Andersen (Northwestern

University) and cryopreserved in the Baer lab. A list of the wild isolates is given in

Supplemental Table S3. The genome sequences of the wild isolates along with collection

information are available at https://caendr.org/.

Pathogen infection and survival assay

We used four strains of bacteria; three documented pathogens (the PA14 strain of

Pseudomonas aeruginosa, the NCTC8325 strain of *Staphylococcus aureus*, and the OG1RF

strain of *Enterococcus faecalis*) and the standard laboratory worm food, the OP50 strain of *E.*

coli. We use the term "pathogen" to refer to the three pathogenic species, and "bacteria" when

OP50 is included, although there is reason to think that OP50 is a weak pathogen of *C. elegans*,

on the grounds that lifespan is extended when worms are fed on killed bacteria compared to live

bacteria (Portal-Celhay et al. 2012). For the MA lines, each bacterial species was assayed in

three or four experimental blocks. Wild isolates were assayed in one assay block per species.

The basic protocol is a variant of the "slow killing assay" (Etienne et al. 2015) , in which ~30 L4-

stage worms are transferred to a plate spread with bacteria and worms scored as live or dead at

12-hr intervals over the course of 120 hrs. The assay is described in detail in section 1 of the

Supplemental Material, and depicted in **Supplemental Figure S3**.

Data Analysis

(*i*) Survival probability. Relative survival of a focal type (replicate *i* of strain *x* of treatment *y* on

bacteria *z*) was quantified as the fraction of worms *f* ($f=n_{LIVE}/n_{TOTAL}$) surviving at time *t* relative to

the mean fraction surviving of the N2 G0 ancestor on OP50, $f_{REL}= f_{FOCAL} / f_{N2,OP50}$. For all

bacteria except PA14, survival was quantified at the culmination of the assay, at *t*=120 hrs.

Survival was much lower on PA14 than on the other strains (**Supplemental Figure S4**), so

survival on PA14 was quantified after 60 hrs. Complete survival data from the MA assay are

given in **Supplemental Table S4**. Times were chosen to roughly maximize the total variance in

survival (**Supplemental Table S5**).

(*ii*) *Mutational parameters*. Two statistics were calculated to quantify the cumulative effects of

mutation accumulation on pathogen susceptibility. First, we estimated the mutational bias (ΔM),

the per-generation change in the trait mean; $\Delta M = (f_{REL,MA}-f_{REL,G0})/t$, where t is the number of

generations of MA (t=250) (Lynch and Walsh 1998).

Second, we measured the mutational variance, V_M, calculated as $(V_{L,MA}-V_{L,0})/2t$, where

$V_{L,MA}$ is the among-line variance of the MA lines, $V_{L,0}$ is the among-line variance of the PS lines,

and t is the number of generations of MA . V_M standardized as a proportion of the residual

(within-line) variance, V_E, is the "mutational heritability"; $h_m^2 = \frac{V_M}{V_E}$. Mutational covariances, Cov_M,

were calculated analogously to V_M, from the among-line components of covariance.

Genotypic variances (V_G) and covariances (Cov_G) of the wild isolates were calculated

similarly to the mutational (co)variances. There is very little residual heterozygosity in the wild

isolates (Rajaei et al. 2021), so wild isolates were treated as inbred lines, in which the genotypic

variance $V_G = V_L$ (Zhang et al. 2021). Genetic covariances and correlations were calculated

analogously. Broad-sense heritability, $H^2 = \frac{V_G}{V_P}$, where V_P is the total phenotypic variance. We

note that there is disagreement within the community of worm quantitative geneticists whether

H^2 should be estimated from V_L or $V_L/2$. The former assumes that the base population is

already inbred; the latter assumes that the base population is outbred (Falconer 1989). The

traditional calculation of H^2 from inbred lines assumes the latter (e.g., the popular software

package H2boot; Arnold and Phillips 1999), but most worm quantitative geneticists use the

former. For the sake of comparison with other studies, we have capitulated and use $H^2 = \frac{V_G}{V_P}$

(note that Etienne et al. (2015) used the traditional $V_L/2$ formulation in their calculation of H^2).

 Statistical Analysis

To account for variation among assay blocks, we first applied the general linear model (GLM)

$y_{ij}= \mu+b_j+e_{ij}$ to each combination of strain (N2 or PB306) and bacterial species independently,

where y_{ij} is the value of f_{REL} of replicate i in assay block j, μ is the overall mean, b_j is the fixed

effect of block j, and e_{ij} is the residual effect, assumed normally distributed with mean 0. The

residuals of the linear model, y^*, are the dependent variable in subsequent analyses. Pooled

over strains, treatments (MA, G0) and bacteria, the y^* are normally distributed (**Supplemental**

Figure S5); we removed outliers with absolute Studentized residuals >3 (n=10/2021 samples).

Next, still considering each combination of *C. elegans* strain (N2 or PB306) and bacterial

species independently, we estimated means and variances from the GLM $y^*_{ijk}=\mu+a_i+L_{j|i}+e_{jk|i}$,

where y^*_{ijk} is the residual from the block mean, μ is the overall mean, a_i is the fixed effect of

generation of MA (gen=0 for MA, gen=250 for MA), $L_{j|i}$ is the random effect of line (MA or

pseudoline) within each treatment group, and $e_{jk|i}$ is the residual effect, assumed normally

distributed with mean 0. Variances of random effects were estimated independently for each

treatment group (banded main diagonal covariance structure). Variances were estimated by

restricted maximum likelihood (REML) with degrees of freedom determined by the method of

Kenward and Roger (1997), as implemented in the MIXED procedure of SAS v.9.4. The fixed

effect of treatment was tested by F-test with Type III sums of squares. Significance of among-

line variance components (V_L) was assessed by likelihood ratio test (LRT) of the model with the

among-line component of variance included against the model with the among-line variance

constrained to 0. Since V_M and h^2_M are composite parameters, confidence intervals were

estimated by means of a parametric bootstrap, assuming that the components ($V_{L,MA}$, $V_{L,PS}$, V_E)

are normally distributed with means equal to the observed and standard deviations equal to the

estimated standard errors. Individual bootstrap estimates of variances were constrained to be

non-negative, i.e., bootstrap replicates less than 0 were set equal to 0.

Genetic covariance matrices (**M** and **G** for MA lines and wild isolates, respectively) were

estimated analogously; the details are given in section 2 of the **Supplemental Material**.

The orientations of **M** and **G** in multivariate trait space were compared using the "genetic

line of least resistance" approach, following Schluter (1996). We describe the method here;

justification is provided in the Results. Data were standardized by the block standard deviation

(i.e., Studentized residuals, y'), to put the four variables on a common scale. **M** and **G** were

used as input data in a principal components analysis (PCA), as implemented in the

PRINCOMP procedure in SAS v. 9.4. The first principal component (PC1) is the axis of

maximum genetic variance, and evolution will proceed most rapidly and easily along this axis,

be it due to selection or drift (Lande 1979; Phillips et al. 2001). PC1 of **G** is commonly referred

to as $\mathbf{g}_{max}$ (Schluter 1996); we follow that convention and refer to PC1 of **M** as $\mathbf{m}_{max}$. The

deviation in orientation in trait space between $\mathbf{m}_{max}$ and $\mathbf{g}_{max}$ is represented by the angle θ

between the two vectors, where $\theta = \cos^{-1}(\mathbf{m}_{max}{}^{T}\mathbf{g}_{max})$ (Schluter 1996).

If $\mathbf{g}_{max} = \mathbf{m}_{max}$, $\theta = 0$. We tested the hypothesis that $\mathbf{g}_{max} = \mathbf{m}_{max}$ as follows. All MA lines

(n=124, pooled over strains) and wild isolates (n=47) were pooled into one dataset, and line

type (MA or WI) was assigned to each line randomly, with 124 random lines being designated

"pseudoMA" and the remaining 47 lines designated "pseudoWI". We re-calculated the among-

line (co)variance matrices and re-ran the PCA as described above; we refer to PC1 of the

randomized data as $\mathbf{m}_{max}'$ **and** $\mathbf{g}_{max}'$, respectively, and the angle between them as θ_{rand}. We

repeated this procedure 1000 times; the fraction of replicates for which $\theta_{rand} > \theta$ is the

approximate P-value of a test of the null hypothesis that $\mathbf{g}_{max} = \mathbf{m}_{max}$.

Results and Discussion

Our overarching purpose is comparative evolutionary genetics in the context of host-pathogen

relationships – what features are specific to particular host genotypes, or particular pathogens,

and what features are general. Survivorship of the N2 G0 progenitor on OP50 (food) sets the

benchmark by which the effects of the pathogenic bacteria are assessed. On average, the N2

progenitor suffered approximately 11% mortality over the course of the 120 hr assay (note that

the assay was conducted at 25°, which is stressfully warm for *C. elegans*). Relative to that

benchmark, survivorship was further reduced by as little as 1.5% in the PB306 G0 progenitor on

OP50 and by as much as 84% in the N2 G0 progenitor on PA14. Averaged over the two strains

(N2 and PB306) and treatment groups (MA and G0), the relative pathogenicity of the three

pathogenic bacteria is OP50<*E. faecalis*<*S. aureus*<<PA14 (**Supplemental Figure S5;**

Supplemental Table S6). Interestingly, the relative pathogenicity was different for the wild

isolates, for which the rank order of pathogenicity was *E. faecalis*<*S. aureus*<OP50<<PA14

(**Supplemental Figure S5**. The discrepancy appears to be due to N2 and PB306 having

substantially higher than average survivorship on *E. coli* OP50. PB306 was included in the wild

isolate assay and its relative survival was similar to the MA assay for all bacterial species except

E. faecalis (albeit n=3 replicates in the WI assay vs. >100 replicates in the MA assay). N2 has

been selected to eat OP50 for thousands of generations, so *E. coli* (or at least OP50) should

probably be considered a pathogen (or at least as "unhealthy food") for all strains except N2. It

is well-known that worms live longer on killed OP50 than on live OP50 (Portal-Celhay et al.

2012), so its status as a pathogen has some precedence.

Mutation Accumulation (MA) lines

(*i*) Means. We expect that survival on OP50 (food, at least for N2) should be unambiguously

beneficial, and both N2 and PB306 exhibit a significant negative mutational bias on OP50

(**Table 1**; **Figure 1A, B**). Predictions about the direction of selection for survivorship on

pathogenic bacteria are less clear, because fitness tradeoffs between pathogen susceptibility

and traits such as growth rate and yield in the absence of

Figure 1. Distribution of line mean relative survival (f_{REL}; see Methods for explanation) on the four bacterial species (left to right: *E. coli* OP50, *Enterococcus faecalis*, *Pseudomonas aeruginosa* PA14, *Staphylococcus aureus*). G0 MA ancestor in orange, MA lines in blue, Wild isolates in green. Red dashed line ($f_{REL}=1$) is the mean survival of the N2 G0 lines on *E. coli* OP50. (A) Top panel; N2; (B) Middle panel, PB306; (C) Bottom panel, wild isolates.

269 **Table 1**. Means (SEM). Survival measures are: *Surv60*=% surviving at 60 hrs; *Surv120*=% surviving at 120 hrs; see Methods for

270 details of calculations of f_{REL} and ΔM. Values of ΔM in bold font are significantly different from 0 (P<0.05); see Methods for details of

271 statistical tests.

Strain	Treatment	Survival	OP50	*E. faecalis*	PA14	*S. aureus*
N2	G0	*Surv60*	0.914 (0.009)	0.869 (0.010)	0.338 (0.016)	0.740 (0.015)
		Surv120	0.888 (0.011)	0.756 (0.012)	0.044 (0.006)	0.607 (0.015)
		f_{REL}	1 (0.012)	0.852 (0.014)	0.370 (0.017)	0.683 (0.017)
	MA	*Surv60*	0.869 (0.013)	0.831 (0.017)	0.340 (0.023)	0.706 (0.014)
		Surv120	0.837 (0.014)	0.678 (0.026)	0.056 (0.010)	0.556 (0.015)
		f_{REL}	0.943 (0.016)	0.764 (0.029)	0.372 (0.025)	0.626 (0.017)
		ΔM ($\times 10^4$)	**-2.3** (0.76)	**-3.6** (0.13)	-1.1 (1.10)	**-2.3** (0.97)
PB306	G0	*Surv60*	0.905 (0.014)	0.759 (0.011)	0.408 (0.053)	0.739 (0.014)
		Surv120	0.875 (0.020)	0.663 (0.014)	0.235 (0.043)	0.632 (0.015)
		f_{REL}	0.985 (0.022)	0.712 (0.016)	0.447 (0.058)	0.711 (0.017)
	MA	*Surv60*	0.857 (0.016)	0.775 (0.013)	0.353 (0.021)	0.679 (0.015)
		Surv120	0.818 (0.019)	0.616 (0.015)	0.155 (0.015)	0.535 (0.018)
		f_{REL}	0.920 (0.021)	0.693 (0.017)	0.386 (0.022)	0.603 (0.020)
		ΔM ($\times 10^4$)	**-2.6** (1.1)	-0.6 (0.96)	-2.5 (1.24)	**-3.8** (1.02)
Average		ΔM ($\times 10^4$)	-2.5	-2.1	-1.8	-3.1
Wild Isolates		*Surv60*	0.788 (0.019)	0.871 (0.019)	0.633 (0.019)	0.819 (0.019)
		Surv120	0.697 (0.017)	0.773 (0.017)	0.051 (0.017)	0.736 (0.017)
		f_{REL}	0.746 (0.023)	0.828 (0.023)	0.665 (0.023)	0.785 (0.023)

pathogens are well-documented (Biere and Antonovics 1996; Tian et al. 2003; Kraaijeveld and

Godfray 2008; Duncan et al. 2011).

Whether the deleterious effects of mutations are exacerbated under stressful conditions

is a topic of active investigation (e.g., Matsuba et al. 2013; Yun and Agrawal 2014; Katju et al.

2018), but there is no theoretical reason to expect them to be unless selection is soft (Agrawal

and Whitlock 2010), which it is not in these assays. Infection by a pathogen is *prima facie*

stressful. If deleterious effects are greater under pathogen stress, we predict that ΔM will be

greater (more negative) in the pathogen treatments than in the OP50 treatment, and that ΔM will

be greatest (most negative) in the most virulent pathogen, PA14.

As a whole, support for those predictions is weak. Point estimates of ΔM are negative in

all eight strain/bacteria combinations (**Table 1**), although not significantly different from zero in

three of the eight, and less on PA14 than on OP50. In only two cases – N2 on *E. faecalis* and

PB306 on *S. aureus* – is ΔM substantially greater (~1.5X) in the pathogen than on OP50.

(*ii*) Variances. To begin, we estimated the among-line component of variance (V_L) for each

combination of bacteria, strain, and treatment (MA or G0). For the MA lines, $V_{L,MA}$ includes the

combined contributions of new mutations and heritable, non-genetic effects; for the G0 PS lines,

$V_{L,PS}$ includes only the heritable, non-genetic effects. $V_{L,MA}$ is significantly greater than zero in

five of the eight strain/pathogen combinations in the MA lines, and nearly significant in a sixth

(PB306 in *E. faecalis*; LRT p<0.07). There was no significant $V_{L,MA}$ on *S. aureus* in either strain

(**Table 2**).

The same analysis for the PS lines has less power than for the MA lines because there

are fewer PS lines than MA lines (N2, n=9,9,39,39 for *PA14*, *OP50*, *E. faecalis*, and *S. aureus*,

respectively; PB306 n=6,9,39,39). For *E. faecalis* and *S. aureus* there are sufficiently many

pseudolines (13/assay block) to credibly test significance of a variance component; in neither

strain is $V_{L,PS}$ significantly different from zero for those two bacteria. For OP50 and PA14, we

Table 2. Variances in relative survival, f_{REL} (SEM). Abbreviations are: V_L (among-line variance x 10^2); V_E (environmental variance x

10^2); V_M (mutational variance x 10^5); h_m^2 (mutational heritability x 10^3; 95% CI in parentheses); $r_{m,w}$ (mutational correlation of relative

survival with competitive fitness); V_G (genetic variance x 10^2); H^2 (broad-sense heritability; 95% CI in parentheses). Values of h_m^2 and

H^2 in bold font are statistically significant (p<0.05).

Strain	Statistic	OP50	*E. faecalis*	PA14	*S. aureus*
N2	V_L (MA) $\times 10^2$	0.64 (0.25)	2.24 (0.87)	1.07 (0.49)	0.28 (0.34)
	V_L (G0) $\times 10^2$	0.042 (0.056)	0	0.17 (0.16)	0.36 (0.30)
	$V_E \times 10^2$	1.17 (0.17)	2.40 (0.42)	2.17 (0.33)	2.81 (0.44)
	$V_M \times 10^5$	1.20	4.47	1.81	0
	$h_m^2 \times 10^3$ (95% CI)	**1.03** (0.18, 2.07)	**1.87** (0.46, 3.71)	0.83 (0, 1.93)	0
PB306	V_L (MA) $\times 10^2$	1.40 (0.46)	0.62 (0.38)	1.20 (0.39)	0.44 (0.42)
	V_L (G0) $\times 10^2$	0.059 (0.160)	0.13 (0.29)	0.11 (0.21)	0
	$V_E \times 10^2$	2.77 0.39)	2.71 (0.43)	2.43 (0.32)	3.18 (0.48)
	$V_M \times 10^5$	2.69	0.99	2.18	0.88
	$h_m^2 \times 10^3$ (95% CI)	**0.97** (0.27, 1.78)	0.37 (0, 0.99)	**0.90** (0.20, 1.63)	0.28 (0, 0.88)
Ave	$V_M \times 10^5$	1.95	2.73	1.99	0.44
	$h_m^2 \times 10^3$	1.00	1.11	0.86	0.14
	$r_{m,w}$	0.05 (0.18)	-0.08 (0.23)	-0.08 (0.19)	0.25 (0.07)
Wild Isolates	$V_G \times 10^2$	1.68 (0.58)	0.57 (0.37)	4.01 (1.22)	2.19 (0.63)
	$V_E \times 10^2$	2.13 (0.37)	2.77 (0.45)	3.75 (0.63)	1.63 (0.27)
	H^2 (95% CI)	**0.44** (0.21, 0.61)	0.17 (0, 0.34)	**0.52** (0.31, 0.66)	**0.57** (0.36, 0.70)

report point estimates of $V_{L,PS}$ and its standard error, and use those estimates in calculations of

V_M, but make no claims about statistical significance of $V_{L,PS}$.

There is significant mutational heritability on two of the four bacterial species in each of

the two strains (**Table 2**). Averaged over all bacterial species except *S. aureus*, for which h_m^2 is

very low or absent, $\bar{h}_m^2 \approx 0.001$, the canonical mutational heritability for a wide variety of traits in

many organisms (Houle et al. 1996; Conradsen et al. 2022). We previously estimated

mutational parameters for survival on PA14 of (almost) the same set of PB306 MA lines

(Etienne et al. 2015). The survival assay (spotted bacterial lawn) and measure of survivorship

in that study (LT50) both differed from this study (spread lawn, relative survival at t_{60}), but the

estimated h_m^2 of LT50 was ≈ 0.001, very close to the estimate from this study. The point

estimate of h_m^2 in N2 in this study is nearly the same as that of PB306, although it does not quite

meet the 5% threshold of significance. ΔM of PB306 on PA14 was near zero in both studies.

(*iii*) Mutational correlations with competitive fitness. Under the MA conditions, the PB306 MA

lines decline in competitive fitness at about 0.1%/generation (ΔM$_w$= -1.13$\pm$0.25x10^{-3}/gen), very

close to the estimate of ΔM$_w$ for N2 in the same assay (-1.09x10^{-3}/gen; Yeh et al. 2018). Pooled

over the two sets of MA lines, mutational correlations (r_M) between f_{REL} on bacterial species *i*

and relative competitive fitness are shown in **Table 2**. For all four bacterial species, r_M between

f_{REL} and competitive fitness in the MA environment is ~0. Evidently mutations (and/or

epimutations) that contribute to relative survival have no consistent cumulative effect on

competitive fitness under the MA conditions.

Wild Isolates

There is significant broad-sense heritability (H^2) for relative survival on three of the four bacterial

species, the exception being on *E. faecalis* (**Figure 1C**; **Table 2**). The wild isolate assay did not

include pseudoline controls for heritable non-genetic variance, so the estimates of H^2 are upper

bounds on the genetic heritability. Surprisingly, H^2 is greatest on *S. aureus*, on which there is

little if any mutational variance. By way of comparison, we recalculated H^2 for survival on PA14

from the set of 114 wild isolates reported by Etienne et al. (2015) with LT50 as the summary

statistic. Mean LT50 was 63.5 hrs. The recalculated estimate, H^2=0.41, is slightly less than the

estimate from relative survival at t_{60} in this study (H^2=0.52).

*Comparison of **M** and **G***

One of our primary goals was to compare the mutational architecture of the two sets of MA lines

(i.e., **M**$_{N2}$ to **M**$_{PB306}$) to better characterize variation in the mutational process in the context of the

multivariate phenotype. We are aware of only one previous such formal comparison, of wing

morphology in *Drosophila melanogaster* (Houle and Fierst 2013). However, the lack of

significant mutational variance in both sets of lines for two of the four bacterial treatments leads

us to think that comparison would not be especially fruitful in this case.

To proceed, we pooled samples over the two sets of MA lines and recalculated among-

line (co)variances as outlined in the methods.   These (co)variances include the heritable non-

genetic component of variance, so we designate the "(epi)mutational" covariance matrix **M*** to

highlight the distinction. Since the assay of the wild isolates was not designed to partition the

among-line (co)variance into genetic and non-genetic components (i.e., we estimated **G*** rather

than **G**), **M*** is the appropriate covariance matrix to compare to **G***. **M*** and **G*** are given in

Tables 3 and 4, respectively.

For **M***, a model with banded main diagonal covariance structure (off-diagonal elements

constrained to 0) fit slightly better than a model with unstructured covariance (ΔAICc=1.2; chi-

square=11.0, df=6, LRT p>0.08). For **G***, the unstructured covariance fit slightly better than the

model with banded main diagonal (ΔAICc=1.5; chi-square=14.2, df=6, LRT p<0.03). Given the

weak support for non-zero off-diagonal elements, it is unsurprising that no consistent pattern

Table 3. (Epi)mutational (co)variance matrix **M***. Covariances shown above the diagonal,

correlations below the diagonal. Values in bold font are significantly different from 0 (p<0.05).

See Methods for details of estimation of **M***.

Bacteria	OP50	*E. faecalis*	PA14	*S. aureus*
OP50	**0.28** (0.08)	0.04 (0.06)	-0.01 (0.05)	0.10 (0.05)
E. faecalis	0.15 (0.23)	**0.23** (0.09)	0.10 (0.06)	0.08 (0.05)
PA14	-0.05 (0.18)	0.42 (0.22)	**0.25** (0.07)	0.05 (0.05)
S. aureus	0.63 (0.35)	0.52 (0.41)	0.36 (0.33)	0.09 (0.07)

356   **Table 4**. (Epi)genetic (co)variance matrix **G***. Covariances shown above the diagonal,

357   correlations below the diagonal. Values in bold font are significantly different from 0 (p<0.05).

358   See Methods for details of estimation of **G***.

Bacteria	OP50	*E. faecalis*	PA14	*S. aureus*
OP50	**0.31** (0.11)	0.10 (0.06)	0.26 (0.12)	-0.05 (0.09)
E. faecalis	0.53 (0.31)	0.11 (0.07)	0.17 (0.10)	0.07 (0.06)
PA14	**0.54** (0.20)	0.60 (0.32)	**0.73** (0.23)	-0.05 (0.12)
S. aureus	-0.13 (0.24)	0.31 (0.29)	-0.10 (0.22)	**0.40** (0.12)

emerges from either the mutational or standing genetic correlations, beyond the observation

that none of the correlations are large and negative.

The first two principal components explain over 90% of the among-line variance in both

M* (60% and 33% for PC1 and PC2) and **G*** (80% and 17% for PC1 and PC2) (**Supplemental**

Table S7). The orientation of **M*** and **G*** as determined from the angle θ between $\mathbf{m_{max}}$ and

$\mathbf{g_{max}}$ does not differ from the random expectation ($\theta = 0.72$, $\bar{\theta}_{Rand} = 0.65$, p>0.35; **Figure 2**),

from which we conclude that mutation alone is a sufficient explanation for the observed

(epi)genetic covariance structure of pathogen susceptibility. The appropriateness of drawing

evolutionary inferences from comparisons of principal components of **G** has been criticized on

the grounds that "...(evolutionarily) causal factors need not have orthogonal effects on the

phenotype" (Houle et al. 2002), but when most of the heritable variance falls along PC1, the

"genetic lines of least resistance" approach seems reasonable.

The Strength and Form of Natural Selection

The finding that $\Delta M < 0$ in every case implies that mutations that reduce survival upon exposure

to pathogenic bacteria are deleterious, on average. That is not surprising in hindsight, although

it is certainly plausible that fitness tradeoffs could result in stabilizing selection on pathogen

susceptibility. The lack of correlation between survival on pathogens with competitive fitness in

the MA environment indicates that pleiotropy between mutational effects in the two contexts is

not strong, but reveals little about selection in the natural environment.

To get a rough idea of how strong selection on these traits is, the neutral expectation

provides a benchmark. For a neutral trait in a predominantly selfing organism such as *C.*

elegans, at mutation-drift equilibrium, $V_G \approx 4N_e V_M$ (Lynch and Hill 1986). N_e in *C. elegans* has

been estimated to be on the order of 10^4-10^5 (Gilbert et al. 2021), so if $V_M \approx 2 \times 10^{-5}$ (the average

for relative survival on OP50), we expect $V_G \approx 0.8$-8, about 40-400 times greater than the

observed value of ~0.02. The discrepancy is similar for PA14, and slightly greater for *E.*

Figure 2. Relative orientation of the first two principal components of the mutational covariance

matrix **M*** (red ellipse) and the standing genetic covariance matrix **G*** (blue ellipse). Solid

colored lines represent PC1 (**m**$_{max}$ and **g**$_{max}$, respectively). Dashed black line depicts the mean

difference between random pseudoMA lines and pseudoWild Isoates ($\bar{\theta}_{Rand}$; see Methods for

explanation). Light blue dashed ellipse shows the 95% confidence limit of the deviation

between pseudoMA lines and pseudoWild Isolates.

faecalis. Although there is considerable uncertainty associated with all of these estimates, the

ability to survive exposure to those three species of bacteria is clearly not a neutral trait;

mutations affecting survival on those bacteria are subject to effective purifying selection. *S.*

aureus may present a different scenario; we return to it below.

402   If selection is sufficiently strong to be approximately deterministic (i.e., ignoring drift), at

mutation-selection balance, $V_G \approx \frac{V_M}{s}$, where *s* represents the average strength of selection

against a mutation affecting the trait (Barton 1990); see Charlesworth (2015) for caveats. By

that calculation, $s \approx 0.001$ for mutations affecting relative survival on OP50, roughly the same on

the virulent pathogen PA14 and a bit stronger on *E. faecalis*. By way of comparison, a similar

calculation for competitive fitness (measured in the MA conditions) revealed $s \approx 0.005$ (Yeh et al.

2018; the value is halved if $V_G = V_L$ rather than $V_L/2$), which is in close agreement with a direct

estimate of the average mutational effect $\bar{s} \approx 0.0035$, calculated by dividing the cumulative

decline in competitive fitness of the MA lines by the average number of mutations carried by an

MA line. These point estimates are obviously rough, but it is evident that mutations that affect

susceptibility to pathogens experience significant selection in nature. Whether the selection is

direct, imposed by exposure to pathogens in nature, or indirect, due to pleiotropic effects on

fitness in the absence of pathogens, cannot be ascertained from these experiments. However,

the absence of a mutational correlation between relative survival and relative fitness (in the MA

context) implies that it might be the former. Recent studies have revealed a complex

transgenerational epigenetically inherited (TEI) behavioral response specific to the PA14 strain

of *P. aeruginosa* that varies among wild isolates (Moore et al. 2021), which reinforces the

inference that PA14 imposes direct selection in nature.

420   Mutation-selection balance (MSB) is not the only mechanism by which genetic variation

may be maintained; balancing selection (BS) is another. There are several well-documented

balanced polymorphisms in *C. elegans* (summarized in Lee et al. (2021)), and there are

numerous examples from diverse organisms in which pathogens do contribute to BS (Fumagalli

et al. 2009; Hawley and Fleischer 2012; Horger et al. 2012; Quéméré et al. 2021). If BS is the

predominant mechanism responsible for the maintenance of genetic variation, there is no

reason to expect V_G and V_M to be correlated. In fact, V_M and V_G are nearly always highly

positively correlated (Houle 1998; McGuigan et al. 2015; Farhadifar et al. 2016). BS and MSB

are often set up as alternative hypotheses for the maintenance of genetic variation, but it is

realistic to expect that uniformly deleterious alleles will always contribute to trait variation,

whereas balanced polymorphisms may or may not. If that scenario is generally true, the effect

of BS would be to weaken the relationship between V_M and V_G, and that does not seem to be

the case. If, in contrast, BS is the exception rather than the rule, the occasional trait

experiencing BS would appear as a high outlier, with more V_G than predicted by its V_M, and

selection as inferred from the ratio of V_M to V_G would appear anomalously weak.

Relative survival on *S. aureus* seems to fit that pattern, at least at first glance. Averaged

over both sets of MA lines, V_M for relative survival on *S. aureus* is only about 20% of V_M on the

other bacteria, whereas V_G is not atypically low, and H^2 is the highest of the four bacteria. If N_e

is closer to 10^4 than 10^5, the selection coefficient inferred from V_M/V_G ($S \approx 2 \times 10^{-4}$) is

approaching effective neutrality ($1/N_e$, the "drift barrier"). However, averaged over the two sets

of MA lines, ΔM is the greatest (most negative) on *S. aureus*, which implies that accumulated

(epi)mutations significantly reduce relative survival on *S. aureus*.

It is not unheard of in the MA literature for trait means to change significantly even in the

absence of significant mutational variance (Shabalina et al. 1997 is a well-cited example in

Drosophila). It is a recurring phenomenon in *C. elegans* MA experiments, and it is most

common for traits that have inherently high variability (e.g., male competitive mating success,

Yeh et al. 2018; metabolite concentration, Johnson et al. 2020). The phenomenon may be

more detectable in *C. elegans* than in other systems, for two related reasons. The ability to

easily maintain cryopreserved stocks permits the contemporaneous assay of the unevolved G0

ancestor and the MA lines, and transgenerational epigenetic inheritance (TEI) is extremely well-

characterized in *C. elegans* (Rechavi and Lev 2017; Beltran et al. 2020). If TEI contributes to

the heritable variance in an unbiased way and of similar magnitude to mutational variance

(Johnson et al. 2020), it could overwhelm the signal of mutational variance even in the presence

of a directional effect of mutations. Alternatively, if a trait is especially sensitive to

microenvironmental noise, V_E may overwhelm V_M such that among-line variance is undetectable

at the given sample sizes. That may be the case for *S. aureus* in the MA lines, for which V_E is

greatest on *S. aureus* in both the N2 and PB306 lines. However, to make things even less

clear, V_E is smallest on *S. aureus* in the wild isolates.   .

Given the time- and labor-intensive properties of both MA experiments and pathogen

survival assays, it is perhaps unsurprising that there are very few studies to have quantified the

effects of spontaneous mutations on pathogen susceptibility. Recently, Mendel et al. (2024)

investigated the effects of spontaneous mutations on the susceptibility of *Drosophila serrata* to

Drosophila C virus (DCV). They reported a mutational heritability for LT50 of $h_m^2 \approx$ 0.004-0.005,

significantly greater than the estimates of h_m^2 in this study and toward the high end of estimates

of h_m^2. The mutational properties (rate and spectrum) at the genomic level in *D. serrata* remain

unpublished, but recent estimates from parent-offspring sequencing of *D. melanogaster* and *D.*

simulans revealed a base substitution/small indel rate 1-2X that of *C. elegans* (Wang et al.

2023). The genomes of those two fly species are somewhere between 36% (*D. simulans*) and

67% (*D. melanogaster*) larger than that of *C. elegans*, so the mutational target is probably

somewhat larger. However, the per-genome, per-generation rate of transposable element

insertions is about 50% greater than that of new base-substitutions/small indels. Based on

limited data, we infer that spontaneous transpositions constitute < 1% of all spontaneous

mutations (Saxena and Baer, unpublished results), so if *D. serrata* has similar TE activity to the

other Drosophila species, that could explain the greater mutational heritability compared to *C.*

elegans. Of course, there are many other differences between flies and worms, and between

bacteria and viruses. Somewhat surprisingly, *C. elegans* is only known to have one viral

pathogen (the Orsay virus, Felix et al. 2011), which has been established as a laboratory model

system (Batachari et al. 2024). The MA lines and wild isolates are freely available and await the

intrepid Ph.D student to do the investigation.

Conclusions

1. The cumulative effect of mutations is to increase susceptibility to pathogens. Moreover,

survival upon exposure to pathogens is uncorrelated with competitive fitness under the MA

conditions. Thus, there is no evidence of tradeoffs between susceptibility to pathogens and

fitness in the absence of pathogens.

2. To the extent that **M*** and **G*** are reliable estimators of **M** and **G**, respectively, it is evident that

survival on these bacteria is subject to effective selection (because **G**$<<$ $2N_e$**M**), but also that **M***

is a sufficient predictor of the orientation of **G*** in multivariate trait space.

3. The number of MA lines included (124), and the number of generations of MA (250, ~1000

days) are quite large. The number of wild isolates (47) is not so large, but is not small.

Nevertheless, the results are (depending on one's perspective) either intriguingly or

distressingly variable between strains (N2, PB306) and bacteria. Nevertheless, the one point of

direct comparison – PB306 on PA14 – yielded results that are very consistent with a previous

study. In our opinion, the way forward lies in development and adoption of high-throughput

phenotyping methods to increase the practical sample sizes in these kinds of assays.

Data Availability

All data are included as Supplemental Tables and will be deposited in Figshare upon

acceptance for publication. All code is available in Figshare at 10.6084/m9.figshare.28046366.

500   **Acknowledgments**

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