Experimental evolution of a pathogen confronted with innate immune memory increases variation in virulence

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Abstract

Understanding the drivers and mechanisms of virulence evolution is still a major goal of evolutionary biologists and epidemiologists. Theory predicts that the way virulence evolves depends on the balance between the benefits and costs it provides to pathogen fitness. Additionally, host responses to infections, such as resistance or tolerance, play a critical role in shaping virulence evolution. But, while the evolution of pathogens has been traditionally studied under the selection pressure of host adaptive immunity, less is known about their evolution when confronted to simpler and less effective forms of immunity such as immune priming. In this study, we used a well-established insect model for immune priming – red flour beetles and their bacterial pathogen *Bacillus thuringiensis tenebrionis* – to test whether this form of innate immune memory favors the evolution of higher virulence. Through controlled experimental evolution of the pathogen in primed versus non-primed hosts, we found no change in average virulence after eight selection cycles in primed host. However, we found a significant increase in the variation of virulence (i.e., host-killing ability) among independent pathogen lines evolved in primed host, and bacteria were unable to evolve resistance against host priming. Whole genome sequencing revealed increased activity in the bacterial mobilome (prophages and plasmids). Expression of the Cry toxin – a well-known virulence factor – was linked to evolved differences in copy number variation of the *cry*-carrying plasmid, though this did not correlate directly with virulence. These findings highlight that innate immune memory can drive variability in pathogen traits, which may favor adaptation to variable environments. This underscores the need to consider pathogen evolution in response to innate immune memory when applying these mechanisms in medicine, aquaculture, pest control, and insect mass production.

Introduction

Virulence is usually defined as the ability of a pathogen to reduce host fitness (or its components, e.g. survival(1,2)) during the infection process (1,2). The direction in which virulence evolves in an evolutionary arms race often depends on the balance between the benefits and costs it provides to the pathogen. For instance, when the highly virulent *Myxoma* virus was introduced into Australian rabbit populations in the 1950s, it evolved to intermediate levels of virulence while maximizing its fitness (3–5). In this example, the cost associated with high virulence was host death, which limited pathogen transmission. Although transmission of pathogens varies and some pathogens can only be transmitted after host death, several studies have shown that optimal transmission rates happen at intermediate levels of virulence (6–8). Additionally, several experimental evolution studies have revealed the complexity of selection processes acting on both hosts and pathogens during their rapid coadaptation (9,10) in which the coevolution between the host and pathogen favours and maintains intermediate levels of pathogen virulence (10).

Host responses and defence mechanisms against pathogens also play a crucial role in the evolution of virulence. Depending on the ability of organisms to resist or tolerate infections, pathogen virulence evolution can take different trajectories (11). The evolution of tolerance, for instance, can strongly influence the evolution of virulence by allowing pathogens to evolve higher transmission rates without increasing the actual level of virulence. As a result, pathogens can become highly virulent in nontolerant hosts and increase the chances of catastrophic epidemics (11). Likewise, epidemiological and mathematical studies propose that "leaky" vaccines, i.e. vaccines that provide incomplete immunity, could enable pathogens to still spread and evolve increased virulence, thus making them more dangerous to unvaccinated individuals (3,12,13) (14). On the other hand, leaky vaccines can also lead to reduced pathogen virulence through the reduction of pathogen load (15). Virulence evolution is a complex trait that may also be influenced by interactions with the hosts' microbiome (16). A recent experimental evolution study in nematodes and the pathogen *Pseudomonas aeruginosa* revealed that

incomplete host immunity caused by host microbiota favours the evolution of higher pathogen

virulence (17). Additionally, external host immunity has also been shown to drive pathogen evolution

towards higher virulence (18). On the other hand, defensive microbes can also reduce pathogen

virulence (19). Considering these seemingly contradictory results, more empirical data from

experimental evolution studies of different host-pathogen systems are necessary to comprehend how

incomplete host immunity influences the virulence of pathogens.

Invertebrate innate immune priming has been suggested to be analogous to "leaky" vaccination, as

primed individuals can still become infected but also can tolerate elevated pathogen loads, providing

an experimental system to assess the effect of innate immune memory on pathogen evolution (20).

Immune priming is a form of immune memory provided by the innate immune system, where initial

non–lethal pathogen exposure can lead to enhanced protection upon secondary exposure (21–23).

Over the last couple of decades, the body of evidence for immune priming in invertebrates has been

constantly growing (24–26) and experimental evolution studies revealed the rapid evolution of

immune priming (27,28). Innate immune systems of vertebrates show similar phenomena, in the form

of trained immunity (29,30). The relevance and potential applications of these forms of innate immune

memory are currently intensively discussed (31–33). While it has been observed that priming reduces

pathogen transmission (34), we still, however, lack empirical data regarding the effects of these innate

forms of immune memory on pathogen evolution and virulence.

In this study, we used the red flour beetle *Tribolium castaneum* and its natural pathogen *Bacillus*

thuringiensis tenebrionis (*Btt*), to investigate whether priming favours the evolution of higher

virulence. *T. castaneum* is a well-established model for host-pathogen coevolution and immune

priming (35–37) and shows a specific priming response towards the spore-forming pathogen *Btt* after

both septic and oral exposure (38–40). *Btt*, widespread and commonly found in diverse environments,

produces insecticidal crystal proteins known as Cry toxins, specifically toxic to different insect orders

(41). After the spores are ingested orally, *Btt* can infect and reproduce within beetle larvae, ultimately

leading to host death (42,43). The oral priming response is triggered by the ingestion of the sterile

filtered supernatant of the spore culture (39), which induces the upregulation of immune genes (44).

To investigate how host immune priming influences the evolution of pathogen virulence, we

experimentally evolved *Btt* in primed and non-primed beetle larvae for eight generations (Fig. 1).

During the experiment, beetles were not allowed to evolve by taking them from a stock population

(i.e., one-sided evolution of the pathogen). After the eight cycles of experimental evolution, we

compared the levels of virulence in *Btt* in primed and non-primed host. We sequenced the genomes

of evolved *Btt* to identify whether genetic changes contributed to evolved differences in virulence. To

our knowledge, our study is the first to show the influence of immune priming on pathogen evolution,

thus broadening our understanding of how host innate immune memory might influence the

trajectories of virulence evolution.

Results

Does host immune priming affect pathogen virulence evolution?

B. thuringiensis tenebrionis (Btt) bacteria evolved experimentally in either immune primed or control (i.e., non-primed) *T. castaneum* hosts (Fig 1, left side). After this selection treatment for eight cycles, we analysed the response of *Btt* to the selection pressure encountered in primed *vs.* control hosts. Specifically, we compared the virulence (measured as host mortality) of eight primed-evolved *Btt* lines (P1-8) and eight control-evolved *Btt* lines (C1-8) as well as ancestral *Btt* (eight replicates directly taken from stock *Btt*, A1-8). These phenotypic readouts (Fig. 1, right side) were taken from two environments, primed and control host, thereby establishing a 'common-garden' experiment.

Fig 1. **Experimental design.** We inoculated one colony of the pathogen *Bacillus thuringiensis tenebrionis* (*Btt*) into 100 mL of *Btt* medium. After 7 days of sporulation, we harvested the spores, adjusted the concentration to 5x10^9, and produced an infection diet that was offered to individual primed and control 15 days old beetle larvae (n=192 per replicate line and cycle). For each selection cycle, we used *Btt* spores from cadavers of *T. castaneum* larvae killed by the infection in the previous cycle and produced eight bacterial lines per selection treatment. For each bacterial line, we isolated spores from five cadavers, which were subsequently grown independently in a sporulation medium to achieve high spore numbers for the infection of a sufficient number of beetle larvae for the next infection cycle. In each selection cycle, bacteria reproduced and sporulated within their host followed by amplification in culture medium. We had eight independently evolving pathogen replicate lines each for evolution in control and primed hosts. After eight selection cycles, we phenotyped and genotyped all evolved pathogen lines (P1-8, C1-8) and the ancestral pathogen. For the ancestral pathogen, we produced pseudo-replicates (A1-8), i.e., lines that had not undergone any selection process, but had grown in separate

flasks to control for possible variations from growth in the medium. For phenotyping, we quantified virulence (proportion of host mortality) and transmission (spore growth in cadavers) in two different host environments (primed and control). We also measured spore growth in a liquid medium. For genotyping, we sequenced the whole genomes of the 24 lines and quantified phage and plasmid coverage. We also measured the expression of a key virulence factor, the Cry-encoding gene, with RT-qPCR.

Fig 2. Virulence (proportion of host mortality) of ancestral *Btt* and the evolved lines Letters denote significant differences (A-ancestral, P-primed evolved, and C-control evolved). Both selection treatments showed reduced mortality compared to ancestral in **A** primed host (glmer: X^2 = 10.287, Df = 2, p = 0.006) and **B** control host (glmer: X^2 = 10.428, Df = 2, p = 0.005), but there was no difference in virulence between the primed and control evolved lines. The plots show the mean and confidence intervals while each individual dot represents the mean virulence (proportion of host mortality) of each independent replicate line. **C**. Correlation between the virulence in primed and control hosts, showing each independent replicate in primed and control beetle larvae. The dotted line shows where the mortality rates of control hosts and primed hosts would be equal. All pathogen lines are below this line, indicating lower pathogen virulence in primed compared to control hosts (i.e., a 'priming effect').

We found that both, the selection treatment (Fig. 1, left side) and the host environment (Fig. 1, right

side) significantly affected pathogen virulence (Fig 2; glmer: X^2 = 74.25, Df = 5, p < 0.001). Selection in

either primed or control hosts significantly reduced the virulence, compared to the ancestral pathogen

(control: p = 0.001; primed: p = 0.006). Regarding the host environment encountered during the

readout, i.e. after evolution had happened, the primed host environment reduced virulence compared

to the control host environment (p < 0.001; Fig 2A and 2B). This shows that priming was still efficient,

also against the lines that had evolved in the primed hosts (Fig. 2 C).

We then performed separate analyses for primed and control host environments to see how individual

bacterial lines vary in virulence in different host environments. In the primed host environment, we

observed a significant effect of the pathogen selection treatment on pathogen virulence (glmer: X^2 =

17.054, Df = 2, p < 0.001; Fig. 2B). Overall, all evolved lines displayed lower virulence in the primed

host environment than the ancestral lines (primed: p = 0.008; control: p < 0.001), while there was no

significant difference between the pathogen lines that had evolved in the control *vs.* primed host (p =

0.840). Additionally, primed evolved replicates showed higher variance in virulence (σ^2 = 0.834) than

control evolved replicates (σ^2 = 0.166). In the control host environment (Fig. 2A), we obtained similar

results, with a significant effect of the pathogen selection treatment on the host mortality with evolved

pathogens showing lower virulence than ancestral (glmer: X^2 = 19.267, Df = 2, p <0.001; primed: p =

0.032; control: p < 0.001), where no differences were detected between the control and primed

selection treatment levels (p = 0.887). As well as in the primed host, we observed higher variance in

virulence across primed-evolved replicates (σ^2 = 0.862) than in control evolved replicates (σ^2 = 0.147).

To further confirm whether such variation in virulence might be associated with the selection

treatment, we modelled the effects of the selection treatments using a Beta distribution in a Bayesian

framework, and allowed the selection treatments to influence the precision parameter (φ). We found

that the variation in virulence (i.e., dispersion host mortality) was greater in primed-evolved *Btt* than

in control-evolved *Btt* with a certainty of 99.06% in the control host environment. This probability

slightly decreased to 90.14% in the primed host environment. Overall, both selection treatments

exhibited an increase in virulence variation compared with the ancestral *Btt* (Fig S1), with a larger

variation observed in the primed-evolved pathogen.

Do the evolved pathogens differ in spore production, sensitivity to priming and fitness?

To evaluate the spore production of the evolved bacteria, we measured the pathogen spore load in

cadavers and the spore production in liquid medium. As *Btt* likely transmits through spores produced

in cadavers (45), we considered the spore load a good proxy of the transmission rate. There was no

difference in spore production between selection treatments in the control host (Fig 3A; lmer: $X^2 =$

0.859, Df = 2, p = 0.650), but in primed hosts, where pathogens that had evolved in primed and control

hosts produced fewer spores than the ancestral *Btt* (Fig 3B;3D; lmer: X^2 = 8.807, Df = 2, p = 0.0122;

primed-evolved: p = 0.014, control-evolved: p = 0.020). We also examined the correlation between

spore load and virulence for all bacteria lines and selection treatments in the primed and control host

environments but found no correlation (Fig S2 and S3). Finally, selection treatment had a significant

effect on spore production in a liquid medium (F = 40.087, Df = 2, p < 0.001), where both primed-

evolved and control-evolved pathogens exhibited lower spore production than the ancestral *Btt* (Fig

3C).

Fig 3. Cadaver spore load in A) primed and B) control hosts. The dots correspond to the mean of ten measurements for each independent pathogen line. The plot shows the mean and the confidence intervals. Ancestral *Btt* produced more spores than both evolved lines, but only in the primed host environment (lmer: X^2 = 8.808, Df = 2, p = 0.012, and there seemed to be weak difference in pathogen load between the evolved lines (p = 0.973). Letters denote significant differences between the selection treatments. **C** Spore growth in the medium. Each dot represents the mean of three independent measures of one pathogen line. The graph shows the mean and confidence intervals. Primed- and control-evolved lines produced fewer spores than the ancestral strain (p < 0.001). **D.** Correlation between the spore load in primed and control hosts, showing the number of spores each independent replicate produces in primed and control beetle cadaver. The dotted line shows where the spore load in control hosts and primed hosts would be equal.

Additionally, we calculated the priming sensitivity of bacterial replicates by dividing virulence in control hosts by virulence in primed hosts. This gave us a value indicating how sensitive the pathogen is to the priming response of the host (Fig 4A). The more sensitive to priming the pathogen is, the lower the virulence in the primed hosts would be. Selection treatment affected the priming sensitivity (X^2 = 5.985, Df= 2, p = 0.050), with the primed-evolved pathogen exhibiting the highest priming sensitivity compared to the ancestral pathogen (p = 0.058).

We also assessed the fitness of each bacterial line within the host by combining mortality rates and spore load data (Fig 4B). Our analysis revealed that both selection treatments and host environment significantly influenced pathogen fitness (X^2= 40.385, df = 5, p < 0.001). The primed host environment

showed a significant negative effect on pathogen fitness (Estimate = -1.03, p = 0.0012), indicating that

the within-host fitness in the priming environment is lower than in the control environment. We finally

tested whether priming sensitivity was correlated with pathogen fitness, and found a negative trend

for the primed-evolved pathogens (Fig 4C).

Figure 4. Priming sensitivity and fitness in the host. Each dot corresponds to bacterial replicate and the box
shows mean and confidence intervals. **A** Priming sensitivity of bacterial replicates was calculated by dividing
the *proportion of mortality in control/the proportion of mortality in primed hosts*. The higher the value, the more
sensitive the pathogen is towards host priming. Selection treatment affected the priming sensitivity (X^2 = 5.985,
Df= 2, p = 0.050), with the primed-evolved pathogen exhibiting the highest priming sensitivity compared to
the ancestral pathogen (p = 0.058). **B** Fitness in the host of each bacterial line was calculated as *mortality*
*proportion X spore load. S*election treatment and host environment significantly influenced host fitness (X^2=
40.385, df = 5, p < 0.001). Pairwise comparisons confirmed that all the selection treatment levels have lower
fitness in the priming host environment. **C** Pearson´s correlation between priming sensitivity and average fitness
in the host (mean of fitness in primed and control host environment). A, C, and P stand for evolution treatment
levels: Ancestral, Control, and Primed evolved pathogen, respectively.

Do the evolved pathogens differ in *Cry* gene expression?

We tested for potential differences in *cry3a* gene expression, as evolved differences in virulence might be related to this important virulence factor. For this, we performed qPCR on RNA extracted from overnight cultures. In contrast to the negative control (genetically modified *Btt* without *Cry3a*), all tested lines expressed *Cry3a* (Supp. Table 2)*.* Overall, the expression in the evolved lines did not differ significantly from that in the ancestral strain (C: estimate= -0.7565, Df= 14, t = -0.597, p = 0.560; P: estimate= -1.1444, Df= 14, t = -0.903, p = 0.382). Four of the control- and three of the primed-evolved lines showed significantly reduced *cry3a* gene expression compared to the ancestral strain (Suppl. Table 3, Fig S4). However, we found no association between *cry3Aa* expression and virulence of the lines, neither in primed (t = 0.279, Df = 15, p = 0.783) nor in control host environment (t = 0.131, Df = 15, p = 0.897)

Do the evolved pathogens differ in their genomes and mobilome?

To investigate whether the observed phenotypes of the evolved *Btt* lines are based on genetic changes, we performed whole genome sequencing of all evolved lines and three replicates of the ancestral pathogen. Compared to the ancestral pathogen, the evolved lines showed a low number of single nucleotide variants (SNVs) and structural variants (SV). Overall, there were 32 identified variants. The primed-evolved lines contained 30 variants and the control-evolved lines contained 28 variants. Only a few of those variants were shared across different replicates and no distinctive pattern could be observed between the primed and control evolution regime, or virulent and non-virulent lines (Fig. 5). Common variants in at least 50% of lines included a G to T substitution at position 2,017,894 in an intergenic region between a hypothetical protein and a putative NTP pyrophosphohydrolase, and a small insertion at position 2,776,026 within a hypothetical protein coding sequence, causing a frameshift (Fig 5). Lines C1 and P8 both had unique variants in the same hypothetical protein DUF4430 at different positions: while C1 had a small deletion at position 3,358,458, P8 had an insertion at

position 3,357,970. Both variants caused frameshifts, with the P8 insertion also introducing a

premature stop codon.

Fig 5. Heatmap displaying the identified single nucleotide variants (SNVs) and structural variants (SVs) at the respective genomic positions. Variants in any line with an alternative allele fraction of >=10% of total read coverage are displayed and alternative allele fraction for each displayed variant was calculated for all remaining lines. Individual lines are ordered within the respective evolution treatment by virulence (*cf.* Fig. 2), from low to high (arrows).

As most bacteria are host to a multitude of mobile genetic elements that can influence the

performance in changing environments, we aimed at identifying differences in such elements between

the evolution regimes. The NCBI reference sequence (46) contains six plasmid contigs where the two

biggest plasmids are 250kb and 188kb, followed by two plasmids of 77kb, 68kb and the smallest 15kb

plasmids. The 44kb plasmid is a phagemid, a circularized phage, which additionally is located on the

chromosome of the reference sequence and thus was not included in the plasmid coverage analysis.

The analysis for plasmid coverage in comparison to chromosome coverage indicated a decrease in the

abundance of the 188kb plasmid in the evolved lines compared to the ancestral lines (Fig 6A). However,

only line P8 retained a plasmid-to-host ratio similar to the ancestral for this plasmid (Fig S4). This

particular plasmid encodes for the important crystal toxin Cry3Aa and other virulence factors (Cry15Aa,

Sphingomyelinase C). Additionally, we observed correlation of the log2 transformed Cry3Aa gene-to-

chromosome ratio and the relative expression of the *cry3Aa* gene (Fig 6B). However, correlation of

other phenotypic traits with active mobile elements showed no statistical significance (Fig S8). The

coverage plots across the genome contained peaks in distinct areas of the chromosome (Fig S5) that

partially overlapped with annotated prophage areas known as *Bacillus* phages (47). We found that the

phages phBC6A52, phi4J1 and phiCM3 had the greatest differences in phage-to-host ratio (Fig 7). We

performed double-layer agar assays (DLAs) for two out of 8 lines of each evolution environment and

the ancestral strain to verify that phages were active in the evolved lines, but not in the ancestral strain.

As expected, we observed visible round plaques in the bacterial lawns of the evolved lines, but not the

ancestral lines (Fig S7).

Fig 6. **A Log2 transformed plasmid coverage divided by chromosome coverage (relative plasmid coverage)**.
Shown are the respective mean values and their standard error. * adjusted p < 0.05 **B** Correlation of the Log2
transformed *cry3Aa* gene coverage divided by chromosome coverage (relative cry3Aa gene coverage) and delta
CT of housekeeping gene expression subtracted by *cry3Aa* gene expression (relative cry3Aa gene expression).

A

B

Fig 7. Heatmap for the Log2 transformed values of phage coverage divided by chromosome coverage for each replicate line. A = ancestral, C = control evolved, P = primed evolved. The lines are ordered by evolution treatment and by virulence where the arrow indicates the increase of virulence. **B** Log2 transformed values of phage coverage divided by chromosome coverage (relative phage coverage). Shown are the respective mean values and their standard error. * adjusted p < 0.05.

Discussion

To investigate the potential effects of immune priming on virulence evolution, we conducted one-sided experimental evolution with *Btt*, using both primed and control *T. castaneum* beetle hosts, and evaluated the phenotype and genotype of the evolved pathogen lines. Although we did not find an increase of virulence in the evolved lines compared to ancestral *Btt* (measured as the proportion of dead hosts), we identified priming as a selective pressure leading to increased variation in virulence (Fig. 2). Some primed-evolved lines showed higher virulence than those of any control-evolved lines and similar to those of ancestral *Btt*, and other primed-evolved lines caused the overall lowest host mortality rates. Another important observation is that the primed-evolved pathogens did not adapt better to the primed host environment. Host priming remained effective and reduced mortality also against the primed-evolved pathogens (Fig. 2C), which were moreover not able to increase spore production in primed hosts (Fig. 3B). This implies that the primed-evolved pathogens failed to develop mechanisms to evade the host's priming immune response. Even the most virulent lines did not overcome the primed immune response, indicating that priming serves as a robust protective measure for the host.

Given that our experimental evolution regimes should favour high virulence (extracting the bacteria from larvae that died by infection), the observed lower virulence in the evolved lines than in the ancestral *Btt* isolates was initially unexpected. However, there are a few possible explanations for this outcome. First, we conducted a one-sided adaptation of *Btt*, rather than coevolution. Previous studies have suggested that high virulence may not be favoured due to its costs when the host is not allowed to co-adapt (10). Second, because certain virulence factors are costly and encoded on mobile elements, bacteria may lose them, particularly when propagated in a medium, as demonstrated in previous research (10,48). Indeed, we found a reduction in the overall coverage of the 188kb plasmid encoding for several virulence factors, as well as differences in the expression of the *cry3Aa* gene that correlated with its genomic coverage (Fig. 7). A reduction in Cry toxin production and plasmid carriage

during passaging experiments has been shown in other studies using *B. thuringiensis* (49). Such losses can be explained by the advantage of "cheating" cells within a population of *Bt* that do not express *cry3Aa* anymore and thereby gain a growth advantage (49). However, in our experiment, *cry3Aa* expression alone did not explain variation in virulence, and the replicate with the lowest virulence (P2) did not change in *cry3Aa* expression compared to the ancestral strain, whereas the replicate with the highest virulence (P7) showed significantly lower *cry3Aa* expression (Fig. S3). This might be due to the generally less important role of the Cry3Aa toxin for *Btt* virulence in *T. castaneum* (49,50), compared to other insects. This finding also aligns with earlier research demonstrating that purified Cry toxin does not induce beetle mortality, but rather a combination of spores and toxins is required for lethal effects (50,51). Potentially, other virulence factors such as chitinases, metalloproteases or vegetative insecticidal proteins have become more relevant in the evolved lines (41,52). This could be especially relevant in primed-evolved lines, because a recent study indicates that Cry toxins are involved in the induction of the priming response (53). Finally, previous studies have shown that evolution can result in reduced virulence due to the interaction with defensive microbes of the host (19) or due to antagonistic interactions among different pathogen strains (54).

Compared to ancestral *Btt*, evolved *Btt* showed reduced sporulation in liquid medium as well as in primed hosts (Fig. 4 and 5). This decline in sporulation may be attributed to the presence of active prophages that we identified in the genomes of the evolved lines (Fig 7). The activation of prophages and the transition from lysogeny (dormant state) to the lytic cycle (active viral replication and bacterial cell lysis) are often triggered by environmental factors (55). Such stress can occur under constant sporulation conditions and is known to cause losses in other *Bacillus* fermentation and to interfere with the evolution of virulence of phages towards their bacterial host (56,57). Interestingly, that a reduction of spore production of evolved lines occurred only in primed but not control hosts (Fig. 4 A,B) could suggest that phage induction is primarily promoted in primed hosts. This environment represents a more stressful niche for the pathogen, as priming leads to the damage in the gut of *T. castaneum* larvae, triggering an early physiological stress response along with the upregulation of

immune related genes (58). Additionally, antagonistic competition may maintain these phages (59).

Bacillus is a member of the microbiome of *T. castaneum* and immune priming can cause a shift towards

higher abundance of *Bacillus* species (60). Releasing lytic phages could provide a competitive

advantage by targeting members of the same genus or species. Alternatively, mutations in phage

binding sites can prevent lysis by lytic phages (61).

While observing changes in phages and plasmids, we could not pinpoint any common chromosomal

mutations that are specific to the selection treatments (Fig. 6). This does not exclude that there might

be more subtle genetic changes that went undetected because they might be specific to certain

replicates. Indeed, the observed increase in variation in virulence indicates that replicates may take

different evolutionary routes (62) making it difficult to find common genetic changes.

Moreover, the lack of a link between spore load and virulence might suggest that *Btt* virulence and

transmission evolve independently. A comparable result was noted in the pathogen *Pseudomonas*

aeruginosa following evolution in hosts with partial immunity (17). Unlike our findings, the incomplete

immunity in that study resulted in increased pathogen virulence. Nevertheless, our results are in line

with the mentioned study that revealed that the most virulent pathogens did not develop mechanisms

to counteract the protective benefits of priming exposure. It is worth noting, however, that in this case,

the priming or partial host immunity was non-specific and triggered by the host's microbiota, while in

our study priming is specific and triggered by the pathogen cues (39,53).

The increased variance in virulence particularly in primed-evolved hosts might be explained by the

more variable niche that a primed host represents for the pathogen (63). *Btt* in primed *T. castaneum*

hosts are not only confronted with an increased intensity and diversity of immune defences (44), but

moreover with a changed microbiome composition (60). More generally, immune-primed hosts

represent a fluctuating environment, known to maintain high levels of phenotypic and genetic

variation in microbial populations, e.g. through balancing selection (64) and pleiotropic effects (65).

Antagonistic pleiotropy might also stem from active phages that could be beneficial in one, but

detrimental to another environment (66). Surprisingly, pathogen lines evolved in primed host showed

more priming sensitivity, i.e., lower virulence (Fig 4 A) in the primed host compared to ancestral and

control evolved pathogen. However, the assessment of virulence was limited to host mortality. It is

possible that the pathogen evolved to prolong the host's survival whilst influencing its development,

which we didn't measure in our study. The observed trend of a negative association between the

fitness of the primed-evolved pathogen and its priming sensitivity suggests that pathogen strains with

higher fitness are less affected by the host's primed immune response (Fig 4). This indicates that fitter

pathogens may have developed strategies to more effectively resist or evade the immunological

effects of host priming. To gain a more comprehensive understanding, the host's response to the

evolved pathogens, particularly in terms of potential downregulation of immune pathways, could be

analysed.

Our observation of increased virulence variation, rather than increased virulence, may also result from

differences in our system compared to models of leaky vaccines, where the pathogen is directly

transmitted between vaccinated hosts (12,14,67). In contrast, *B. thuringiensis*, both in our experiment

as well as in nature, experiences two distinct environments, as it can thrive outside of its insect hosts

(68). In addition to host factors, the longevity of spores and their ability to persist in the environment

may also play a role in determining virulence (69).

In conclusion, our results are to our knowledge the first to show an effect of specific immune priming

on the evolution of bacterial virulence. The observation of increased variance in virulence is

evolutionary relevant because selection needs variation to act upon. Therefore, host immune priming

and other forms of incomplete, innate immune memory, may boost the evolutionary potential of

pathogens and even lead to multiple evolutionary stable states (11). This is important because innate

immune memory bears a large potential for applications. Not only the medical relevance of trained

immunity in medicine, but also applications of invertebrate immune priming in fields such as

aquaculture and insect mass production are gaining momentum (70,71). Studies in further systems

with immune priming and trained immunity are needed to better understand and potentially manage

the evolution of pathogens when confronted with innate immune memory.

Materials and methods

Model organisms

As pathogen, we used *Bacillus thuringiensis* subspecies *morrisoni* biovar *tenebrionis* (*Btt*) from the

Bacillus Genetic Stock Centre (BGSCID), Ohio State University, USA) under code 4AA1. The strain was

stored in 500 μL microtubes containing 25% glycerol at -80°C. This strain has been shown to cause

mortality in *T. castaneum* larvae upon the oral ingestion of spores and their toxins (42,43).

As the host, we used a wild-type *T. castaneum* (Cro1) population collected from Croatia in 2010 (42)

and adapted to laboratory conditions for eight years in numerous non-overlapping generations until

the start of the experiment in 2018. Beetles were reared on organic wheat flour (Bio Weizenmehl Type

550, DM-drogerie Markt GmbH + Co. KG) supplemented with 5% brewer's yeast (the flour mixture was

heat-sterilised at 75°C for 24 h). For rearing and infection experiments, beetles were kept at 30°C, 70%

relative humidity, and a 12-h/12-h-light–dark cycle.

Experimental evolution

Preparation of bacteria for selection treatment and infections

We prepared spore cultures and supernatants as previously (39,40), with minor modifications. To

begin the selection experiment, we plated *Btt* vegetative cells from the -80°C glycerol stock on an LB

agar plate and incubated it overnight at 30°C. The following day, we inoculated 5 ml of Bt medium

(w/V–0.75% Bacto Peptone (Sigma), 0.1% glucose, 0.34% KH2PO4, 0.435% K2HPO4) that was

supplemented with 25 μl of a sterile salt solution (0.2 M MgSO$_4$, two mM MnSO$_4$, 17 mM ZnSO$_4$, and

26 mM FeSO$_4$) and 6.25 µl of sterile 1 M CaCl$_2$ × 2H$_2$O with a single colony, and incubated it at 30°C and

200 rpm overnight. Overnight cultures were moved to 1-L Erlenmeyer flasks containing 300ml of Bt

medium supplemented with 1.5 mL of salt solution and 375 µL 1 M CaCl$_2$ × 2H$_2$O. Since *Btt* is naturally

neomycin-resistant, 15µg/µL of neomycin were added to the sporulation media. This step was taken

to guarantee that other spore-forming bacteria that could come from cadavers did not proliferate in

the medium. The cultures were kept at 30°C and rotated at 180 rpm for seven days in the dark. On the

third day of incubation, 1.5 mL of salt solution and 375 µL of 1 M CaCl$_2$ × 2H$_2$O were added to the

cultures. Following seven days of sporulation, the culture was spun for 20 minutes at 4500 rpm. The

pellet was washed and redispersed in PBS (Calbiochem). Using means of flow cytometry (FACSCanto

II; Becton Dickinson, USA) as described in Milutinović et al. 2015, spore concentration was adjusted to

5 x 10^9 /mL, and the diet was prepared by adding heat-sterilized flour with yeast (0.15 g/mL of spore

suspension; (42,43). We then pipetted the liquid spore-containing diet (10 µL per well) into the 96 flat

bottom microtiter plates (Sarstedt, Germany) under sterile conditions, covered them with a breathable

sealing foil for culture plates (Kisker Biotech), and dried them overnight (~17 hours) at 30°C. For all

remaining selection cycle experiments, where spores were taken from cadavers, we slightly adjusted

the protocol; after processing spores, we inoculated 50 mL of Bt medium supplemented with salts and

neomycin in a 100 ml Erlenmeyer flask, followed by seven days incubation at 180 rpm and 30°C.

Preparation of primed and control host for the selection treatment

The primed host environments were performed as previously described (39,40). *Btt* vegetative cells

were plated from a -80°C glycerol stock on LB agar plates and incubated overnight at 30°C. Five single

colonies were incubated overnight, and sporulation cultures were produced as described above but

without adding neomycin. After seven days of sporulation and centrifugation, the collected

supernatants were centrifuged twice at 4,500 rpm for 15 min and then filter-sterilized with a 0.45-µm

pore-size, followed by a 0.2-µm pore-size cellulose acetate filter (Whatman GmbH). We mixed every

millilitre of spore-free supernatant with 0.15 g heat-treated flour (supplemented with 5% yeast). Next,

ten microliter of the priming diet mixture was pipetted into each well of a 96-well plate (Sarstedt,

Germany), sealed with a breathable foil (Kisker Biotech), and dried overnight at 30°C. For the control

treatment, the diet was prepared by mixing 0.15 g heat-treated, yeast-supplemented flour with each

millilitre of sterile non-conditioned growth medium.

For every selection treatment cycle, we took approximately 2,000 1-month-old adults and left them to

lay eggs for 24 hours on the flour. The priming protocol was as follows: 15 days after oviposition, 192

larvae per replicate line were individually placed on a previously prepared priming diet in two 96-well

flat-bottom plates. The plates were sealed with clear tape and punctured with needles to allow air

circulation and were stored under standard rearing conditions. After 24h on a priming diet, we

transferred larvae to flour discs consisting of phosphate-buffered saline (PBS) mixed with 0.15 g/mL

flour, in which they stayed for four days until we transferred them again to pathogen spore-containing

discs.

Re-infection for another selection treatment cycle

We measured mortality for seven days and then collected larvae that died during the first two days,

placed them in the dark, and waited for another seven days to ensure that all *Btt* cells in the cadaver

had sporulated. For reinfection, we isolated spores from insect cadavers and propagated them once in

liquid sporulation media to obtain a sufficient number of spores for infection. For each bacterial line,

we isolated spores from five cadavers, which were subsequently grown independently in a growth

medium to achieve high spore numbers for the infection of a sufficient number of beetle larvae for the

next infection cycle. Spores from the five cultures derived from the cadavers were combined for

infection to reduce the effect of potential clonal interference among rapidly growing clones in culture.

We randomly selected five cadavers per individual line and individually homogenized them as

described before (43). We then counted spore concentrations using flow cytometry and pasteurized

the suspensions at 80°C for 10 minutes (i.e., any possible vegetative cells are killed). Liquid medium

was inoculated with 1mL of 1x10^6 spores derived from cadavers for germination, followed by spore

production. After seven days of incubation, we pooled all five individually raised spore cultures and

prepared the spore-flour diet. Primed and control host larvae were then placed on spore flour discs,

and their survival was tracked.

Phenotypic readouts

Virulence and spore production of evolved bacterial lines

All the selection treatment lines were introduced to a primed and control host environment. Host

mortality was screened for six days. We performed this experiment in three experimental blocks, with

48 larvae per treatment/block, resulting in 144 larvae per treatment combination. For every block, we

took approximately 2,000 1-month-old adults and left them to lay eggs on the flour (implemented with

5% yeast) for 24 hours, following the priming protocol described above. For each block, we prepared

12 96-well plates with a priming diet and 12 96-well plates with control diet. We placed 15-day-old

larvae on the plates and, after 24 hours, transferred them to PBS flour discs. After four days, we

transferred them to 96-well plates containing spores from 24 bacterial lines (eight Primed evolved,

Control evolved and Ancestral). Each 96-well plate contained spores from all 24 bacterial lines to

control for the possible plate effect. This resulted in 24 96-well plates for each experimental block.

Spores from the bacterial lines were prepared as described above. Frozen spores from the 8th passage

and ancestral *Btt* were first plated on an agar plate and after 14 hours 10 colonies were inoculated in

50 mL growth medium supplemented with salts. After seven days of bacterial growth and sporulation,

spores were washed, and measured via flow cytometry, and the concentration was adjusted to

$5x10^9$cells/mL. We used measures of the numbers of spores in a liquid medium from three

experimental blocks to test for the differences in spore production between the lines.

We quantified the pathogen load of infected larvae that died from the infection. For this, we took ten

cadavers per bacterial line, which resulted in a total of 240 cadaver samples. The samples were

processed as described before (43). Each cadaver was placed in a 100 µl Eppendorf tube with 50 µL

PBS and a sterile metal bead. The cadavers were homogenized using a Mixer Mill MM301 (Retsch) to

release spores into the solution. The solution was then washed using a cell strainer to remove larger

particles. The solution was diluted 1:1000 and measured via flow cytometry.

Cry gene expression in Btt cultures

Before RNA extraction, we inoculated two cultures of ancestral and evolved lines in 6 mL LB medium

at 30°C, 180 rpm. Following 12 h of incubation, the cultures were centrifuged at 4500 rpm for 10 min

and resuspended in 100 µL of PBS. We added 2.2 µl of lysozyme (100 mg/mL) to 100 µL of the bacterial

suspension and incubated for 30 min at 37°C with shaking. Subsequently, we implemented an RNA

extraction protocol with a combination of TRIzol reagent (Ambion, USA) lysis and extraction with the

purification via spin columns from the SV Total RNA Isolation System (Promega) as described before

(72)**.** To analyze *cry3a* expression, cDNA synthesis, and qPCR were performed in two identically

designed blocks. We conducted reverse transcription of cDNA using the RevertAid First Strand cDNA

Synthesis Kit (ThermoScientific) following the manufacturer's instructions. For the reaction, we used

2µL of extracted RNA, irrespective of its concentration, along with a random hexamer primer. The

resulting cDNA product was then diluted 1:5 with nuclease-free water.

Following cDNA synthesis, we performed qPCR. This analysis involved measuring the expression of

cry3a relative to two housekeeping genes (*yqey* and *rps21*) (Supp. Table 1). **Error! Reference source**

not found. Prior to the main experiment, we determined the amplification efficiency of the primers by

qPCR using a serial 10-fold dilution of the initial cDNA input (Suppl. Table 1).

Genome analysis

Sequencing

After culturing samples from all evolved and three ancestral replicate lines on LB agar plates, we extracted DNA using the Monarch Genomic DNA Purification Kit (New England Biolabs GmbH, Frankfurt am Main, Germany). For sequencing on the PacBio Sequel IIe platform (Pacific Biosciences Inc., Menlo Park, CA, USA), we created a library with the SMRTbell Express Template Prep Kit 2.0 (Pacific Biosciences Inc.) according to the manufacturer's recommendations, aiming for at least 20x coverage on the HiFi read level. We then mapped the resulting HiFi reads against the sequence of a *Btt* reference strain (NCBI accession no.: GCF_022810725.1) using the SMRT Link software version 11.

Variant detection

To call single nucleotide variants (SNVs) we used GATK Haplotypecaller (v4.3.0.0) (73) and Google's DeepVariant (74) on the Galaxy Server (Galaxy Version 1.5.0+galaxy1). We then intersected the resulting vcf files of the evolved lines with those derived from the ancestral lines using the bcftools isec function (version 1.16) (75) to filter out variants that were based on differences between our ancestral lines and the used reference sequence. Of the remaining SNVs we kept those with an alternative allele fraction of >= 10% of the total read coverage at the respective position. To call structural variants (SVs) we used Sniffles2 (Version 2.2) (76), The resulting vcf files were intersected in the same way as the SNV vcf files. Next, we verified all final variants by manually inspecting them in the Integrative Genomics Viewer (IGV) (77). To annotate the SNVs we used SnpEff (version 5.2c (build 2024-04-09 12:24)) (78) using the prokaryotic codon table and the reference genbank- and nucleotide fasta files.

Plasmid coverage

We calculated the coverage across the chromosome and the different reference plasmids respectively using samtools depth (version 1.16.1) (75) and estimated the abundance of plasmids by dividing each plasmids mean coverage by the chromosomes mean coverage and taking the log2 of those values. We termed the resulting values "Log2(Plasmid-to-Host ratio)".

Phage analysis

To detect potentially complete prophage genomes within our reference genome we used PHASTER (79). Next, we calculated the coverage across the genome using samtools depth and divided the genome into sections of a) outside of prophage regions and b) within prophage regions. We considered b) to be true if the region contained all essential phage genes necessary for its reproduction (from small terminase to lysis cassette). Finally, we divided the means of b) by the means of a) and took the log2 of those ratios. We termed the resulting values "Log2(Phage-to-Host ratio)". The log2 transformed Phage-to-Host ratios were then plotted using ggplot2 in Rstudio. To confirm the presence of active phages in the evolved lines, we performed double-layer agar assays (DLA). In short, we grew bacteria in 20 ml Luria broth (LB) medium overnight. The next morning, we adjusted the overnight culture to an optical density of 1 at 600 nm wavelength (OD_{600} of 1) and inoculated 500 µL into 50 ml fresh LB medium. After four hours, 8 hours, and 12 hours, we took 1 ml of the culture and adjusted it to an OD_{600} of 1.5. We then mixed 400 µl of the culture with 4 ml lukewarm 0.7% LB agar and poured the mix onto a solid 1.5% LB agar plate. After 24 h, we checked the plates for plaques in the bacterial lawn.

Statistics

All data were analysed using R (80) and R Studio (81). To test whether the pathogen selection treatment and host environment have an effect on the virulence, we tested the effect of selection treatment (primed-evolved, control-evolved and ancestral) on the host survival by fitting a generalized linear mixed effect model with interaction (GLMM) and binomial error distribution using *the glmer*

function in the package *lme4* (82) (mortality~ selection_treatment*host environment)

+(1|Replicate_line)+ (1|Block)+(1|Plate). The selection treatment level and host environment were

defined as fixed factors, whereas a bacterial line, plate and block were defined as random factor. As

interaction was not significant, we then fitted a model with selection treatment as a fixed factor in

primed and control host environments separately. To check whether the random effect for Replicate

line shows more variation in one selection treatment compared to the other, we fitted a random slope

model where the slope of Selection_treatment is allowed to vary across Replicate_line: (mortality ~

Selection_treatment + (1 | Replicate_line) + (1 | Block) + (1 | Plate). After fitting a model, we examined

the variance estimates for replicate line within selection treatment. To further test of the treatments

influence the variation in mortality, we ran a Bayesian Beta regression model in rstan (Stan

Development Team 2023) where we allowed the precision parameter (ϕ) to be influenced by the

Selection treatment. Then, we used the posterior samples of the model to estimate the probability

that the variation in mortality arising from primed, control and ancestral populations are different.

For spore load (transmission) analysis, the *lmer* function in the *lme4* package was used with the

bacterial line as a random factor (spore_load~ selection_treatment +(1|replicate_line) in primed and

control host environment separately. For spore growth analysis we performed linear regression

analysis with *lm* function. After fitting the models, the variance between the selection treatment

(primed-evolved, control-evolved, ancestral) was assessed using one-way analysis of variance. The

means were compared using Tukey's post-hoc analysis with Benjamini-Hochberg correction.

Priming sensitivity of each bacterial replicate was determined by dividing proportion of mortality in

control host with proportion of mortality in primed host. Kruskal-Wallis test was used to assess

differences in priming sensitivity between the selection treatment levels. The means were compared

using post-hoc pairwise comparisons using Dunn's test with a Benjamini-Hochberg correction.

We determined fitness in the host by combining spore load and proportion of mortality per individual

bacterial line. Then we compared it between selection treatment levels using *lmer* function

(Fitness_in_host ~ Line * Environment + (1 | Replicate_line). The means were compared using Tukey's

post-hoc analysis with Benjamini-Hochberg correction.

To analyze gene expression data from qPCRs, we first compared the relative expression of *cry3a* of the

primed and control-evolved *Btt* to the expression in the ancestral *Btt* cultures in a linear mixed model

with experimental block and evolved replicates as random factors (relative expression ~

selection_treatment + (1|block) + (1|replicate)) using the packages *lme4* and lmerTests (83). We also

fitted a second linear mixed effects model using the same packages to compare the evolved lines to

the ancestral cultures (relative expression ~ selection_treatment + (1|block)).

We used various statistical techniques to evaluate the relationships among different variables. The

connections between spore load and mortality, mortality and cry expression, and average host fitness

with priming sensitivity were analyzed using Pearson's correlation method. We conducted Mann-

Whitney tests using the wilcox.test() function to evaluate the significance of differences between log2

transformed Plasmid-to-Host and Phage-to-Host ratios. The p.adjust() function with the "holm"

adjustment method was applied to address multiple testing concerns. A linear model was constructed

to investigate significant interactions between the read coverage for the cry3Aa gene and its

expression. The cor() function from the R Stats package (version 4.3.1) was used to determine the

correlation strength between different phenotypic readouts and active mobile elements for both

primed evolved lines and control evolved lines. To statistically compare the correlation strengths of

various readouts between primed and control evolved lines, we utilized the cocor() package (version

1.1-4) (84).

Data availability

The genomic sequences can be found on NCBI under project number: PRJNA1174726

(https://dataview.ncbi.nlm.nih.gov/object/PRJNA1174726?reviewer=cmkqpsc4dunlettul69l9ta251).

It will be made publicly available upon acceptance.

All the data sets, code and RMarkdown file with figures and analysis results can be found: Korša, A., Baur, M., Schulz, N., Anaya-Rojas, J. M., Mellmann, A., & Kurtz, J. (2024). Experimental evolution of a pathogen confronted with innate immune memory increases variation in virulence [Data set]. Zenodo. https://doi.org/10.5281/zenodo.14223031

Authors contribution

AK and JK conceptualisation. AK, MB, NS and AM data curation and investigation. AK, MB, NS, and JAR visualization and formal analysis. AK, MB and JK writing – original draft preparation. AK, MB, NS, JAR, JK, and AM writing – review & editing. JK supervision and funding acquisition.

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