Between virus correlations in the outcome of infection across host species:

evidence of virus genotype by host species interactions

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Abstract

Virus host shifts are a major source of outbreaks and emerging infectious diseases, and predicting

the outcome of novel host and virus interactions remains a key challenge for virus research. The

evolutionary relationships between host species can explain variation in transmission rates,

virulence, and virus community composition between hosts, but the potential for different viruses to

interact with host species effects has yet to be established. Here, we measure correlations in viral

load of four *Cripavirus* isolates across experimental infections of 45 *Drosophilidae* host species. We

find positive correlations between every pair of viruses tested, suggesting that broadly susceptible

host clades could act as reservoirs and donors for certain types of viruses. Additionally, we find

evidence of genotype-by-genotype interactions between viruses and host species, highlighting the

importance of both host and virus traits in determining the outcome of virus host shifts. More

closely related viruses tended to be more strongly correlated, providing tentative evidence that virus

evolutionary relatedness may be a useful proxy for determining the likelihood of novel virus

emergence, which warrants further research.

Impact Summary

Many new infectious diseases are caused by viruses jumping into novel host species. Estimating the probability that jumps will occur, what the characteristics of new viruses will be, and how they are likely to evolve after jumping to new host species are major challenges. To solve these challenges, we require a detailed understanding of the interactions between different viruses and hosts, or metrics that can capture some of the variation in these interactions. Previous studies have shown that the evolutionary relationships between host species can be used to predict traits of infections in different hosts, including transmission rates and the damage caused by infection. However, the potential for different viruses to influence the patterns of these host species effects has yet to be determined. Here, we use four viruses of insects in experimental infections across 45 different fruit fly host species to begin to answer this question. We find similarities in the patterns of replication and persistence between all four viruses, suggesting susceptible groups of related hosts could act as reservoirs and donors for certain types of virus. However, we also find evidence that different virus genotypes interact in different ways with some host species. Viruses that were more closely related tended to behave in similar ways, and so we suggest that virus evolutionary relatedness may prove to be a useful metric for predicting the traits of novel infections and should be explored further in future studies.

Introduction

Virus host shifts, where viruses jump to and establish onward transmission in novel host species, are

a major source of outbreaks and emerging infectious diseases [1–3]. Many human infections,

including HIV, Ebola, and recently SARS-CoV-2, have shifted into humans from other species, and

continue to cause significant damage to public health, society, and the global economy [4–7].

Predicting and preventing virus host shifts have consequently become major goals of virus research

[8]. Many challenges remain in achieving these goals, including improving our understanding of the

host, virus, and ecological factors that influence the outcome of initial cross-species transmission

[9,10], and the evolutionary and epidemiological factors that determine which pathogens become

established in novel hosts [11].

Several studies have investigated how host evolutionary relatedness can explain variation in the

outcome of infection across host species. Greater phylogenetic distance between the natural

(donor) and recipient hosts is associated with decreased likelihood of cross-species transmission

[12,13] and reduced onward transmission within the novel host species [14]. Phylogenetic distance

between hosts also explains variation in virulence after cross-species transmission, which increases

when viruses jump between more distantly related hosts [14–16]. Groups of closely related hosts

have also been shown to share similar levels of susceptibility to novel viruses, independent of the

distance to the natural host [17,18], and harbour similar virus communities [19–21].

Less is known about the interactions that exist between virus genotype and host species in

determining the outcome of infection [18]. Within species, genotype-by-genotype interactions

between host and virus can be important determinants of the outcome of infection. These

interactions alter the rank-order of host susceptibility and so reduce the strength of correlations

between viruses [22]. Comparative analyses of fungal pathogens of plants [23] and ectoparasites of

mammals [24] have revealed parasite phylogenetic effects, and some evidence exists to suggest

similar effects may be found in viruses. Closely related viruses tend to infect the same broad host

taxa [25], despite high levels of geographic range overlap between potential hosts [26], suggesting

they share similar constraints on their host ranges. Both co-speciation and the preferential host

switching of viruses can support this assumption, given that viruses are overwhelmingly likely to

encounter other host taxa over the timescales required for speciation. That said, shifts between

divergent host species are also common across every virus family [27] and these exceptions include

several human zoonoses of major concern [14].

The extent to which closely related viruses share traits has yet to be established. However, inferring

the characteristics of viruses from better studied relatives is common and sometimes necessary. This

is frequently the case during the early stages of outbreaks, where primary research on new viruses

or variants is not available. When SARS-CoV-2 first emerged, its characteristics and epidemiological

trajectory were inferred from closely related zoonotic and endemic coronaviruses [28], and from

other pandemic respiratory viruses such as influenza A [29]. Comparisons to previous outbreaks

were used to parameterise disease models in the 2009 H1N1 pandemic [30,31], the 2014 Ebolavirus

outbreak [32], and in forecast models of seasonal influenza [33]. Even for viruses that are not newly

emerged, many experimental models of infection rely on surrogates when the virus of interest is

unavailable, non-permissive in cell culture or animal models, or requires considerable adaptation to

experimental hosts [34,35].

These comparisons assume that the traits of one virus are similar to other, related viruses. However,

comparisons between more distantly related viruses, such as bat and canine rabies viruses [44] and

diverged lineages of influenza viruses [45–47], found stark differences across larger evolutionary

scales. Many examples also exist of small genetic changes having large phenotypic effects in viruses,

including single SNP changes altering the host range of canine parvoviruses [36], the vector

specificity of Chikungunya virus [37], and the infectivity of naturally occurring Ebolaviruses [38]. Only

three amino acid substitutions are required to switch receptor specificity of avian H7N9 influenza

from poultry to human cell receptors [39]. Virus evolution is often characterised by high mutation

rates and frequent reassortment and recombination [40–42]. This, alongside an incomplete sampling

of extant viruses [43], has left many poorly resolved evolutionary relationships between and within

existing virus lineages [44]. Given these complications, it remains an open question whether

comparisons between related viruses can produce accurate inferences of infection traits.

In this study, we have investigated how patterns of host susceptibility (the ability of a virus to persist

and replicate) are correlated between viruses, using experimental infections with four *Cripavirus*

isolates (family *Dicistroviridae*) across a panel of 45 host species of *Drosophilidae*. Three of the

viruses are isolates of Drosophila C virus (DCV-C, DCV-EB and DCV-M), a well-studied virus isolated

from *Drosophila melanogaster* [45]. The fourth virus is the closely related Cricket Paralysis virus

(CrPV) which was isolated from Australian field crickets (*Teleogryllus commodus*) and is a widely

used model insect pathogen [46–48]. Both DCV and CrPV infect a broad range of insect taxa [49],

cause virulent infections in adult flies [17,50], and share similar mechanisms for co-opting the host

translation machinery [51]. A major-effect resistance gene called *pastrel* increases resistance to DCV

in *D. melanogaster* [52–54] and has also been shown to provide cross-resistance to CrPV along with

another gene, *Ubc-E2H* [55]. Both DCV and CrPV are targeted by the host antiviral RNAi pathway and

each encodes a potent suppressor of antiviral RNAi. However, these suppressors have different

functions and target different components of the RNAi pathway [56,57]. DCV and CrPV also differ in

their tissue pathology; DCV has been shown to infect gut tissues, causing intestinal obstruction

following septic inoculation in *D. melanogaster*, which was not observed in CrPV infection [58].

Although little is known about the differences between DCV isolates, they have been shown to cause

similar levels of virulence in *D. melanogaster* and interact differently with the endosymbiont

Wolbachia [59].

Previous work in this host system has shown that susceptibility to DCV-C varies across host species,

with the host phylogeny explaining a large proportion of the variance in both viral load and virulence

[17]. The host phylogeny is also an important determinant of the evolution of DCV-C in novel hosts,

with evidence that mutations that adapt the virus to one host may also adapt it to closely related

host species. This suggests virus genotype could alter the likelihood of host shifts in *Drosophila* [60].

Here, we measure correlations in the ability of four viruses to replicate and persist across host

species and provide evidence of both broad similarities in infection outcome and differences

consistent with virus by host species interactions.

Materials & Methods

Fly Stocks

Flies were taken from laboratory stocks of 45 different species of *Drosophilidae* (S1 Text Table A).

Before experiments began all included stocks were confirmed to be negative for infection with DCV

and CrPV by quantitative reverse transcription PCR (qRT-PCR, described below). Stocks were

maintained in multi-generation *Drosophila* stock bottles (Fisherbrand) at 22°C, in a 12-hour light-

dark cycle. Each bottle contained 50ml of one of four varieties of food media (S1 Text

Supplementary Methods).

Host Phylogeny

The method used to infer the host phylogeny has been described in detail elsewhere [17]. Briefly,

publicly available sequences of the *28S, Adh, Amyrel, COI, COII, RpL32,* and *SOD* genes were collected

from Genbank (see https://doi.org/10.6084/m9.figshare.13079366.v1 for a full breakdown of genes

and accessions by species). Gene sequences were aligned in Geneious v9.1.8

(https://www.geneious.com) using a progressive pairwise global alignment algorithm with free end

gaps and a 70% similarity IUB cost matrix. Gap open penalties, gap extension penalties, and

refinement iterations were kept as default.

Phylogenetic reconstruction was performed using BEAST v1.10.4 [61] as the subsequent

phylogenetic mixed model (see below) requires a tree with the same root-tip distances for all taxa.

Genes were partitioned into separate ribosomal (*28S*), mitochondrial (*COI*, *COII*), and nuclear (*Adh*,

Amyrel, RpL32, SOD) groups. The mitochondrial and nuclear groups were further partitioned into

groups for codon position 1+2 and codon position 3, with unlinked substitution rates and base

frequencies across codon positions. Each group was fitted to separate relaxed uncorrelated

lognormal molecular clock models using random starting trees and 4-category gamma-distributed

HKY substitution models. The BEAST analysis was run twice, with 1 billion MCMC generations

sampled every 100,000 iterations, using a birth-death process tree-shape prior. Model trace files

were evaluated for chain convergence, sampling, and autocorrelation using Tracer v1.7.1 [62]. A

maximum clade credibility tree was inferred from the posterior sample with a 10% burn-in. The

reconstructed tree was visualised using ggtree v2.0.4 [63].

Virus Isolates

Virus stocks were kindly provided by Julien Martinez (DCV isolates) [59], and Valérie Dorey and

Maria Carla Saleh (CrPV) [57]. DCV-C, DCV-EB and DCV-M were originally isolated from fly stocks

with origins in three separate continents; DCV-C and DCV-EB were isolated from lab stocks

established by wild capture in Charolles, France and Ellis Beach, Australia respectively, while DCV-M

was isolated directly from wild flies in Marrakesh, Morocco [45]. The CrPV isolate was collected from

Teleogryllus commodus in Victoria, Australia [64]. Virus stocks were diluted in Ringers solution [65]

to equalise the relative amounts of viral RNA and checked for contamination with CrPV (DCV

isolates) and DCV (CrPV isolate) by qRT-PCR as described below.

Virus phylogeny

Full genome sequences for DCV-C (*MK645242*), DCV-EB (*MK645239*), DCV-M (*MK645243*), and CrPV

(*NC_003924*) were retrieved from the NCBI Nucleotide database. Annotations of ORFs for the

replicase polyprotein (CrPV: *Q9IJX4*, DCV: *O36966*) and structural polyprotein (CrPV: *P13418*, DCV:

O36967) were collected from the UniProtKB database and used to separate the coding and non-

coding regions of each virus. ORF sequences were concatenated and aligned using the Geneious

progressive pairwise translation alignment algorithm with a Blosum50 cost matrix and default

parameters. Alignments were manually checked for quality and sequences aligning to CrPV ORF1

nucleotides 1-387 and 2704-2728 were removed due to the presence of large indels.

Phylogenetic reconstruction was performed using BEAST v1.10.4 with translated ORF sequences

fitted to an uncorrelated relaxed lognormal molecular clock model using a speciation birth-death

process tree-shape prior. A Blosum62 substitution model [66] with a gamma distribution of rate

variation with four categories and a proportion of invariable sites was used. The model was run for

10 million MCMC generations sampled every 1,000 iterations and evaluated in Tracer v1.7.1 as

above, and a maximum clade credibility tree inferred with a 10% burn-in.

Inoculation

Before inoculation, 0-1 day old male flies were kept in vials containing cornmeal media (S1 Text

Supplementary Methods), and were transferred to fresh media every 2 days for one week. Vials

contained between 5 and 20 flies (mean = 14.5) and were kept at 22°C at 70% relative humidity in a

12-hour light-dark cycle. Flies were inoculated at 7-8 days old under CO_2 anaesthesia via septic pin

prick with 12.5μm diameter stainless steel needles (Fine Science Tools, CA, USA). These needles

were bent approximately 250μm from the end to provide a depth stop and dipped in virus solution

before being pricked into the pleural suture of each fly. Inoculation by this method has been shown

to follow the same course as oral infection but is less stochastic [67]. Inoculated flies were then snap

frozen immediately in liquid nitrogen, providing a 0 days-post-infection (dpi) timepoint, or

maintained in cornmeal vials for a further 2 days ± 3 hours before freezing, providing a 2 dpi time

point. Within replicate blocks the 0 and 2 dpi vials for each virus were inoculated on the same day,

and together constituted one biological replicate. We aimed to collect three biological replicates for

each species and virus combination, with the order of species, vial (0 or 2 dpi), and virus randomised

for each replicate block.

Measuring Change in Viral Load

To measure the change in viral load between 0 and 2 dpi, total RNA was extracted from flies

homogenised in Trizol (Invitrogen, supplied by ThermoFisher) using chloroform-isoproponyl

extraction, and reverse transcribed using Promega GoScript reverse transcriptase (Sigma) with

random hexamer primers. qRT-PCR was carried out on 1:10 diluted cDNA on an Applied Biosystems

StepOnePlus system using Sensifast Hi-Rox Sybr kit (Bioline). Cycle conditions were as follows: initial

denaturation at 95°C for 120 seconds, then 40 cycles of 95°C for 5 seconds, and 60°C for 30 seconds.

DCV isolates were measured using the same primer pair (Forward: 5'-GACACTGCCTTTGATTAG-3',

Reverse: 5'-CCCTCTGGGAACTAAATG-3') which targeted a conserved location and had similarly high

efficiencies across all isolates. For CrPV the following primers were used: Forward: 5'-

TTGGCGTGGTAGTATGCGTAT-3', Reverse: 5'-TGTTCCGTCCTGCGTCTC-3'. *RpL32* housekeeping gene

primers varied by species (S1 Text Tables B & C). For each sample, two technical replicates were

performed for each amplicon (viral and *RpL32*).

Between-plate variation in C_T values was estimated and corrected for using a linear model with plate

ID and biological replicate ID as parameters, as described elsewhere [68,69]. Mean viral C_T values

from technical replicate pairs were normalised to *RpL32* and converted to fold-change in viral load

using the $2^{-\Delta\Delta C_T}$ method, where $\Delta C_T = C_{T:Virus} - C_{T:Rpl32}$, and $\Delta\Delta C_T = \Delta C_{T:day0} - \Delta C_{T:day2}$.

Amplification of the correct products was verified by melt curve analysis. Repeated failure to amplify

product, the presence of melt curve contaminants, or departures from the melt curve peaks of

positive samples (± 1.5°C for viral amplicons, ± 3°C for *Rpl32*) in either the 0 or 2 dpi samples were

used as exclusion criteria for biological replicates. In total, of the 180 unique combinations of host

species and virus measured, 3 biological replicates were obtained for 161 combinations, 2 replicates

for 18 combinations, and 1 replicate for 1 combination (*Drosophila virilis,* CrPV).

Statistical Analysis

Phylogenetic generalised linear mixed models were used to investigate the effects of host

relatedness on viral load, and to examine correlations between the different virus isolates.

Multivariate models were fitted using the R package MCMCglmm [70] with the viral load of each

virus isolate as the response variable. The structures of the models were as follows:

$$y_{hiv} = \beta_{1:v} + \mu_{p:hv} + \mu_{s:hv} + e_{hiv} \ (1)$$

$$y_{hiv} = \beta_{1:v} + \mu_{p:hv} + e_{hiv} \ (2)$$

In these models, y_{hiv} is the change in viral load for virus v in the ith biological replicate of host species

h. The fixed effect β_1 represents the intercepts for each virus isolate, the random effect μ_p represents

the effects of the host phylogeny assuming a Brownian motion model of evolution, and e represents

the model residuals. Model (1) also includes a species-specific random effect that is independent of

the host phylogeny ($\mu_{s:hv}$). This explicitly estimates the non-phylogenetic component of between-

species variance and allows the proportion of variance explained by the host phylogeny to be

calculated. $\mu_{s:hv}$ was removed from model (2) as model (1) struggled to separate the phylogenetic

and species-specific traits. Wing size, measured as the length of the IV longitudinal vein from the tip

of the proximal segment to the join of the distal segment with vein V [71], provided a proxy for body

size [72] and was included in a further model as a fixed effect ($wingsize_h.\beta_2$).

Within each of these models the random effects and residuals were assumed to follow a

multivariate normal distribution with a centred mean of zero and a covariance structure of $V_p \otimes A$ for

the phylogenetic effects, $V_s \otimes I$ for species-specific effects, and $V_e \otimes I$ residuals, where $\otimes$ represents

the Kronecker product. A represents the host phylogenetic relatedness matrix, I an identity matrix

and V represents 4x4 covariance matrices describing the between-species variances and covariances

of changes in viral load for the different viruses. Specifically, the matrices V_p and V_s describe the

phylogenetic and non-phylogenetic between-species variances in viral load for each virus and the

covariances between them, while the residual covariance matrix V_e describes within-species variance

that includes both true within-species effects and measurement errors. Since each biological

replicate was tested with a single virus isolate, the covariances of V_e cannot be estimated and were

set to zero.

Models were run for 13 million MCMC generations, sampled every 5,000 iterations with a burn-in of

3 million generations. Parameter expanded priors were placed on the covariance matrices, resulting

in multivariate F distributions with marginal variance distributions scaled by 1,000. Inverse-gamma

priors were placed on the residual variances, with a shape and scale equal to 0.002. To ensure the

model outputs were robust to changes in prior distribution, models were also fitted with flat and

inverse-Wishart priors, which gave qualitatively similar results.

The proportion of the between species variance that can be explained by the phylogeny was

calculated from model (1) using the equation $v_p/(v_p + v_s)$, where v_p and v_s represent the phylogenetic

and species-specific components of between-species variance [73], and is equivalent to phylogenetic

heritability or Pagel's lambda [74,75]. The repeatability of viral load measurements was calculated

from model (2) as $v_p/(v_p + v_e)$, where V_e is the residual variance of the model [76]. Inter-specific

correlations in viral load were calculated from model (2) V_p matrix as $cov_{x,y}/\sqrt{(var_x + var_y)}$. If

correlations between viruses are close to one (with no change in the variance whilst the means

remain constant) it would suggest there are no host species-by-virus interactions [22]. Parameter

estimates reported are means of the posterior density, and 95% credible intervals (95% CI) were

taken to be the 95% highest posterior density intervals.

The data files and R scripts used in this study are available in an online repository:

https://doi.org/10.6084/m9.figshare.13750711.v1.

287   Results

Change in viral load is a repeatable trait among host species

To investigate similarities between related viruses in the outcome of infection across host species, as

well as the potential for virus genotype to interact with host species effects, we experimentally

infected 45 species of *Drosophilidae* with four virus isolates: DCV-C, DCV-EB, DCV-M, and CrPV. The

DCV isolates formed a distinct clade (>93% genome and ORF amino acid identity, with 265-556 SNPs

between isolates), with the closest relationship between DCV-C and DCV-EB. CrPV formed an

outgroup to the DCV isolates (57-59% identity, with over 4000 SNPs between CrPV and each DCV

isolate, Fig. 1, Table 1). In total, 15,657 flies were inoculated, and the change in viral load after two

days of infection was determined by qRT-PCR (Fig. 2). The mean viral load within host species ranged

from an approximately 2.7-billion-fold increase in *Drosophila persimilis* infected with DCV-M to a 2.5-

fold decrease in *Zaprionus tuberculatus* infected with DCV-C. Viral loads across host species tended

to be higher for the DCV isolates, with a mean fold-increase of roughly 11,000 - 19,000, and lower

for CrPV, with a mean fold-increase of roughly 1,600.

302
303 **Figure 1: Phylogeny of virus isolates.** Evolutionary relationships estimated from open reading frame (ORF)
304 amino acid sequences presented in a midpoint-rooted tree. Node labels represent the posterior probabilities
305 of each clade, and the scale bar represents amino acid substitutions per site.

	DCV-C	DCV-EB	DCV-M	CrPV
DCV-C		97.10%	94.00%	57.40%
DCV-EB	98.81%		93.90%	57.30%
DCV-M	98.30%	98.20%		57.40%
CrPV	59.00%	58.70%	58.70%	

Table 1: Virus isolate sequence similarity. Percentage sequence identity was calculated from multiple-
alignment of whole genome nucleotides (white) or concatenated amino acid sequences of ORFs 1 & 2 (grey).
Approximately 92 SNPs and 28 amino acid substitutions exist for every 1% of sequence divergence.

Phylogenetic generalised linear mixed models were fitted to the data to determine the proportion of

variation in viral load explained by the host phylogeny (Table 2). The phylogeny explained 79% of the

variation in viral load for CrPV but only 9-21% of the variation for the DCV isolates, with wide

credible intervals on all the DCV estimates. This was due to the model struggling to separate

phylogenetic and species-specific effects for these viruses. The repeatability of viral load across host

species was high for both CrPV (0.66) and the DCV isolates (0.92-0.96), with the between-species

phylogenetic component (v_p) explaining a high proportion of the variation in viral load with little

within-species variation or measurement error (v_r). We found no significant effect of wing length (a

proxy for host body size) on viral load for any of the included viruses, with all estimates having

credible intervals overlapping zero.

Figure 2: Change in viral load across a diverse panel of *Drosophilidae* host species for different virus isolates. Bar height and colour show the mean change in viral load at 2 dpi on a $\log_2$ scale, with error bars representing the standard error of the mean. The phylogeny of *Drosophilidae* hosts is presented on the left, with the scale bar representing the number of nucleotide substitutions per site and scale axis representing the approximate age since divergence in millions of years (my) based on estimates from [77,78]. The virus cladogram, presented at the top, is based on the evolutionary relationships shown in Fig. 1.

Virus	Mean change in viral load (log$_2$)	Repeatability	Variance explained by phylogeny
DCV-C	13.50 (95% CI: 11.17, 15.89)	0.96 (95% CI: 0.93, 0.98)	0.11 (95% CI: 0, 0.35)
DCV-EB	14.22 (95% CI: 11.42, 16.76)	0.96 (95% CI: 0.93, 0.98)	0.09 (95% CI: 0, 0.32)
DCV-M	13.63 (95% CI: 10.52, 16.59)	0.92 (95% CI: 0.87, 0.96)	0.23 (95% CI: 0, 0.51)
CrPV	10.66 (95% CI: 8.59, 12.66)	0.66 (95% CI: 0.46, 0.83)	0.79 (95% CI: 0.50, 1.00)

331   **Table 2: Estimates of mean change in viral load, repeatability, and the proportion of variation explained by**
332   **the host phylogeny.** Estimates of the mean change in viral load and repeatability are taken from model (2),
333   while estimates of the variation explained by the host phylogeny are taken from model (1).
334

Correlations between viruses are consistent with virus by host species interactions

Inter-specific correlations in viral load between viruses were then estimated from the variance-

covariance matrices of model (2) (Fig. 3A). We found strong positive correlations between the DCV

isolates (r > 0.93), with the strongest correlation between DCV-C and DCV-EB (r = 0.97). Correlations

between DCV isolates and the more distantly related CrPV were positive (r = 0.52-0.59) but weaker

than the correlations between the DCV isolates. The fact the DCV:CrPV correlations (and their 95%

CI's) are not close to one is consistent with virus genotype by host species interactions on viral load

[22]. This is further demonstrated by the notable differences in the rank order of host species

susceptibility for each virus (Fig. 3B), equivalent to a crossing over of reaction norms for the

susceptibility of host species between virus genotypes [79].

DCV-C was more strongly correlated to DCV-EB than to DCV-M (Δr = 0.04, 95% CI = 0.00, 0.09), and

more strongly correlated to DCV-M than to CrPV (Δr = 0.40, 95% CI = 0.18, 0.82), consistent with an

increase in the strength of correlation between viruses with closer evolutionary relatedness. Point

estimates imply a similar pattern for DCV-EB, but the evidence for a stronger correlation with DCV-C

than DCV-M was not well supported (S1 Text Table D).

352
Figure 3: Similarities in infection outcome across host species and interactions between virus genotype and
host species. A) Correlations in viral load between virus isolates. Individual points represent the mean change
in viral load after 2 dpi for each host species on a $\log_2$ scale, and trendlines have been added from a univariate
least-squares linear model for illustrative purposes. Correlations (r) are the total inter-specific correlations and

95% CIs from the output of model (2). B) Differences in the rank-order of host species susceptibility between
virus isolates. Bar height and colour show the mean change in viral load at 2 dpi on a $\log_2$ scale, with error bars
representing the standard error of the mean. The order of species along the x-axis has been sorted in
ascending order of viral load during infection with DCV-C. Deviations from this rank-order of host species
susceptibility for other viruses is indicative of crossing reaction norms and interactions between virus genotype
and host species. The virus cladogram is based on the evolutionary relationships shown in Fig. 1.
363

Discussion

Closely related host species present similar environments to novel viruses [73,80], and so tend to

share similar levels of susceptibility to a given virus [12–18]. Likewise, closely related viruses are

often assumed to share characteristics that make their host interactions, transmission, and

evolutionary trajectories comparable [28–35]. Here, we measured the strength of correlations in

viral load between four *Cripavirus* isolates across 45 host species of *Drosophilidae*, to look for

similarities between related viruses as well as evidence of virus genotype by host species

interactions on the outcome of infection. We found positive correlations between every pair of

viruses tested, indicating broad similarities in the outcome of infection across host species, but also

evidence for interactions between virus genotype and host species with changes in the rank order of

host species susceptibility between the different viruses (Fig. 3). This highlights the importance of

considering both host and virus traits in understanding the outcomes of virus host shifts.

The strong positive correlations between DCV isolates are likely due to relatively high levels of

sequence conservation resulting in only small differences in their ability to infect different host

species. However, in other viruses a small number of mutations have been shown to allow successful

infections in novel hosts [36,39]. We find a few instances of such effects here. For example, in

Zaprionus davidi, DCV-EB shows a decline in viral load, suggesting it is failing to replicate and persist

in this host species, whereas the other isolates show an increase in viral load in the same host.

Similarly, *Scaptodrosophila pattersoni* is amongst the least susceptible to DCV-M but has relatively

high viral loads for the other virus isolates. A greater number of these effects can be seen when

comparing hosts infected with DCV isolates to those infected with CrPV, where multiple species have

markedly different susceptibilities depending on the virus infecting them. For example, both

Drosophila ananassae and *Drosophila strutevanti* are within the five most susceptible species to

DCV-C, but also the eight least susceptible to CrPV. The weaker correlations that exist between DCV

and CrPV may be due to interactions with different host traits that vary in their patterns across the

host phylogeny. CrPV and DCV are known to have distinct methods of suppression of the host

antiviral RNAi pathway [56,57] and cause pathology in different tissues [62]. Additionally, their

relatively high levels of sequence divergence (57-59% identity) may have resulted in changes in the

ability of each virus to bind to host cell receptors, utilise host replication machinery, or avoid host

immune defences [81].

The existence of correlations between viruses suggests that host susceptibility is not specific to

individual viruses and that certain host clades may be broadly susceptible to infection. Host species

that are permissive to multiple virus genotypes may allow for the persistence of increased genetic

diversity in the virus population, allowing viruses to generate and maintain mutations that make

them more likely to emerge in novel host species [82,83]. They also have the potential to act as

"mixing vessels", providing increased opportunities for virus reassortment and recombination [84],

which has been proposed as a possible route for several viruses to acquire pandemic potential

[85,86]. Broadly susceptible host clades may therefore act as common reservoirs and donors of

emerging infectious diseases and identifying them in relevant systems could inform control and

prevention strategies [87].

The differences in correlation strength between pairs of viruses tended to follow differences in their

evolutionary divergence, such that more closely related pairs of viruses were more strongly

correlated in the outcome of infection across host species. This provides some tentative evidence

that the ability of a virus to infect a novel host may be inferred based on its evolutionary relatedness

to other viruses, although we lack the sample size or diversity of viruses needed to test this

conclusively. Numerous examples exist where a small number of genetic changes in viruses cause

large phenotypic differences [36–39], which would be exceptions to any link between correlation

strength and evolutionary relatedness [87]. Nevertheless, virus phylogenetic effects may still prove

to be a useful proxy for determining the likelihood of novel virus emergence. Further work is now

needed to expand the findings of this study to broader groups of viruses, and to test the importance

of the virus phylogeny in determining the potential outcomes of virus host shifts.

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