Opinion

Dissecting the Genetic Architecture of Host–Pathogen Specificity

Louis Lambrechts*

Department of Virology, Institut Pasteur, Paris, France

In this essay, I argue that unraveling the full genetic architecture (i.e., the number, position, effect, and interactions among genes underlying phenotypic variation) and molecular landscape of host–pathogen interactions can only be achieved by accounting for their genetic specificity. Indeed, the outcome of host–pathogen interactions often depends on the specific pairing of host and pathogen genotypes [1]. In such cases, the infection phenotype does not merely result from additive effects of host and pathogen genotypes, but also from a specific interaction between the two genomes (Box 1). This specific component, which can be measured by the interaction term in a two-way statistical analysis of phenotypic variation as a function of host and pathogen genotypes, is referred to as a genotype-by-genotype (G×G) interaction [1]. By analogy to genotype-by-environment (G×E) interactions that occur when different genotypes respond differently to environmental change, G×G interactions occur when the response of host genotypes differs across pathogen genotypes. Although the concept of G×G interactions has mostly been used by evolutionary ecologists to describe the specificity of host immune defenses against pathogens [2], it can be applied to any phenotype resulting from the specific interaction between two genomes. The general definition of G×G interactions allows its use to characterize phenotypes ranging from macroscopic traits such as lifespan [3] to the level of gene expression [4]. Here, the genetic specificity of host–pathogen associations is defined in the sense of G×G interactions. This definition differs from that of immunological specificity, which is the ability of a host to recognize and mount an immune response against a particular pathogen genotype or antigen. Whereas immunological specificity often depends on infection history (i.e., past exposure to a pathogen), genetic specificity describes the intrinsic compatibility between host and pathogen genotypes and occurs independently of infection history.

In some instances, the specificity of host–pathogen associations can be ex-plained to a large extent by major genes of hosts and pathogens, as in the gene-for-gene model of plant–pathogen compati-bility [5,6]. In general, however, multiple genes and epistatic interactions among these genes determine the infection out-come [7–9]. A recent meta-analysis of 500 published studies reporting quantitative trait loci (QTL) for host resistance to pathogens in plants and animals revealed that the genetic architecture of this trait varies dramatically across different combi-nations of host and pathogen genotypes [9]. Thus, different host–pathogen associ-ations involve different QTL and epistatic interactions, indicating that a substantial portion of phenotypic variation derives from the specific interaction between the two genomes. This is made even more complex when multiple pathogen species or strains infect the same host [10] and/or when G×G interactions are environment-dependent [11,12].

It is striking that, to date, quantitative genetic studies of host–pathogen systems have neglected the specific component of the interaction. Dissecting the genetic architecture of complex infection traits has traditionally relied on QTL mapping strategies [7,9] and more recently on association analyses of candidate gene polymorphisms [8]. A major caveat of these QTL mapping and association studies is that they focus on either the host or the pathogen genome. Because they consider variation in only one of the two interacting organisms, these studies ignore specific host genome by pathogen genome interactions. In order to fully dissect the genetic architecture and ex-plore the molecular landscape of host–pathogen interactions, it will be necessary to account for the specific component of the relationship. This should be made possible by recent developments in molec-ular strategies combining host and patho-gen genetics [13–15] and in quantitative genetic models of host–pathogen interac-tions allowing detection of host QTL by pathogen QTL interactions [16,17]. Ad-vantage could also be taken from existing methods for analysis of gene–gene and gene–environment interactions [18–21]. A critical (and limiting) aspect for investigat-ing genetic specificity is the need to include different combinations of host and pathogen genotypes in the experi-mental design.

From a fundamental standpoint, im-proved knowledge of the genetic architec-ture of host–pathogen specificity has important implications for our under-standing of the ecology and evolution of host–pathogen associations. The genetic specificity of host–pathogen interactions is thought to promote the maintenance of host and pathogen genetic diversity via frequency-dependent coevolutionary cy-cles [22–25], which in turn favor higher rates of mutation, recombination, and sexual reproduction [26]. Unraveling the genetic architecture and molecular land-scape of host–pathogen specificity, com-bined with molecular evolution analyses, will shed light on the mechanistic basis of the infection process and the biochemistry of host–pathogen recognition [27–30]. The genetic model and precise epistatic interactions underlying host–pathogen specificity are critical determinants of

Box 1. A Quantitative Genetic Model of Host–Pathogen Interactions

Quantitative genetics is the area of genetics dealing with the inheritance of traits showing continuous phenotypic variation [35]. Typically, quantitative phenotypes are modeled as the result of combined effects of the genes (G) and the environment (E). The basic model to describe the phenotype of an individual is:

$$y = \mu + g + \varepsilon \qquad \text{(Equation 1.1)}$$

where y is the phenotypic value of the individual, μ is the mean value of the population, g is the genetic contribution to the deviation from the mean (usually termed "genotypic value"), and ε is the environmental (non-genetic) deviation. By extending this model to a quantitative trait resulting from the interaction between a host and a pathogen, the model becomes:

$$y = \mu + g_H + g_P + g_{HP} + \varepsilon \qquad \text{(Equation 1.2)}$$

where g_H is the host genotypic value, g_P is the pathogen genotypic value, and g_{HP} is the genotypic value due to the specific G×G interaction. This simple model ignores interactions between genes and environment (G×E and G×G×E effects), which occur when genotypic values vary across environments. The genetic component of phenotypic variance in a host–pathogen interaction can thus be partitioned into three distinct terms: variance due to the additive effect of the host genotype, variance due to the additive effect of the pathogen genotype, and variance due to the specific interaction between the two genomes. Whereas the first two terms can be characterized by considering either the host or the pathogen genetic variation alone, exploring the genetic basis of host–pathogen specificity requires that genetic variations in both the host and the pathogen are considered simultaneously.

In the case of a trait determined by two haploid loci i and j of a single organism, we can define α_i the additive effect of locus i, α_j the additive effect of locus j, and β_{ij} the interaction effect between loci i and j to decompose the genotypic value into:

$$g_{ij} = \alpha_i + \alpha_j + \beta_{ij} \qquad \text{(Equation 2.1)}$$

Whereas non-additive interactions effects between different loci are defined under the term "epistasis", when i and j are two homologous alleles of the same diploid locus the interaction effect is generally referred to as "dominance". By defining $\Sigma\alpha$ as the sum of all additive effects and $\Sigma\beta$ as the sum of all interaction effects (both within and between loci), the previous equation can be generalized to any trait determined by multiple (n>2) loci:

$$g = \Sigma\alpha + \Sigma\beta \qquad \text{(Equation 2.2)}$$

By incorporating this expression into the general quantitative genetic model given by equation 1.1, we obtain the expression:

$$y = \mu + \Sigma\alpha + \Sigma\beta + \varepsilon \qquad \text{(Equation 1.3)}$$

Likewise, using the quantitative genetic model of host–pathogen interactions given by equation 1.2, it follows:

$$y = \mu + (\Sigma\alpha_H + \Sigma\beta_H) + (\Sigma\alpha_P + \Sigma\beta_P) + g_{HP} + \varepsilon \qquad \text{(Equation 1.4)}$$

By using the notations $\Sigma\alpha_{HP} = \Sigma\alpha_H + \Sigma\alpha_P$ (sum of additive effects of host and pathogen loci) and $\Sigma\beta_{HP} = \Sigma\beta_H + \Sigma\beta_P + g_{HP}$ (sum of interaction effects between host loci, between pathogen loci, and specific G×G interactions between host and pathogen loci), the equation becomes:

$$y = \mu + \Sigma\alpha_{HP} + \Sigma\beta_{HP} + \varepsilon \qquad \text{(Equation 1.5)}$$

The striking similarity between equations 1.5 and 1.3 illustrates how the phenotype of a host–pathogen interaction can simply be modeled as that of a third organism that combines both genomes. In such a model, the specific G×G interaction is included among all interaction terms, supporting the view that considering specificity in the genetic architecture of host–pathogen interactions is as important as including intra-genome epistasis. Like epistasis [36,37], host–pathogen specificity may thus largely contribute to the unexplained genetic variation in susceptibility to infectious diseases missed by conventional QTL mapping strategies or genome-wide association studies [38,39].

coevolutionary dynamics and the evolution and maintenance of sex and recombination [27,31]. In conjunction with gene flow and genetic drift, the genetic basis of specificity can also influence the spatial structure and local adaptation of host and pathogen populations [32].

From a more applied perspective, exploring the genetic basis of host–pathogen specificity will provide important insights into the mechanisms of disease emergence. Pathogens with a broad host range (i.e., a low degree of host specificity) are those most likely to emerge or re-emerge following ecological

changes [33]. Disease emergence can also result from pathogen adaptation to a novel host species or population, which largely depends on the initial compatibility between host and pathogen genotypes [34]. Characterizing the genetic and molecular basis underlying host–pathogen specificity thus holds considerable promise for understanding, predicting, and preventing disease emergence. It will help to identify host species and populations most at risk for emergence of a given pathogen and uncover new molecular targets to interfere with the ability of emerging pathogens to jump from one host to another.

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