

1 **Shaping variation in the human immune system**

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8 Immune responses demonstrate a high-level of intra-species variation, compensating
9 for the specialization capacity of pathogens. The recent advent of high-depth immune
10 phenotyping projects in large-scale cohorts has allowed a first look into the factors
11 that shape the inter-individual diversity of the human immune system. Genetic
12 approaches have identified genetic diversity as drivers of 20-40% of the variation
13 between the immune systems of individuals. The remaining 60-80% is shaped by
14 intrinsic factors, with age being the predominant factor, and environmental influences,
15 with cohabitation and chronic viral infections identified as key mediators. Here we
16 review and integrate the recent high-depth large-scale studies on human immune
17 diversity, with its potential impact on health.

18

19 **Dissecting Variation Through Population Immunology Studies**

20 The evolutionary arms-race between pathogens and hosts includes the constant
21 refinement of host anti-pathogen mechanisms, and the reciprocal development of
22 microbial strategies to circumvent these defenses [1]. While multi-cellular hosts bear
23 the advantage of being able to specialize cells (our immune system) to the task of
24 pathogen clearance, micro-organisms have the undoubted advantage in evolutionary
25 speed. Because of this, the inter-individual diversity of the immune system in a
26 species is an important mechanism for limiting the impact specific pathogens can
27 have on the morbidity and mortality of a population. The maintenance of inter-
28 individual diversity within the immune system is therefore a critical aspect of its
29 function, albeit one that is often neglected in immunological research, for example,
30 through the use of genetically identical mice housed under specific pathogen free
31 conditions [3].

32 The archetype of diversity generation in the human immune system is the production
33 of extra-genomic variation in the antigen receptors of the adaptive immune system (T
34 cell receptor and B cell receptor). The combination of permutational choice in gene
35 fragment selection with an error-prone system creates an essentially personal T cell
36 and B cell receptor repertoire for each individual, with response capacity further
37 molded by the structural diversity encoded by the highly polymorphic MHC locus [4,
38 5]. Clonal selection further shapes the repertoire in the periphery. However, even
39 beyond the process of adaptive diversity, an enormous amount of variation exists in
40 the number and functional status of each immunological component. Leukocytes,
41 one of the key cellular mediators of the immune system, are epigenetically modified
42 by the micro-environment and a myriad of other interactions to the point where each
43 individual cell could be considered to exist in a unique status [6, 7]. These
44 interactions shape the composition of an individual's immune system over their
45 lifespan, for example driving the differentiation of naïve T cells into subsets with
46 specialized functions, resulting in the context-dependent generation of Th1, Th2,
47 Th17, follicular helper and regulatory subsets [8].

48 While a major focus of immunology research over the past decades has been in
49 identifying and functionally dissecting these alternative activation states, there has
50 been a growing appreciation that the plasticity of the immune system and the
51 plethora of subsets and activation statuses results in dramatic individual-to-individual

52 variation in the immune system. In this review we concentrate on population
53 immunology (see Glossary) studies, which we define as research which directly
54 compares the status of the immune system (relative number of leukocyte subsets,
55 activation status of leukocytes, production of non-cellular mediators) across a large
56 number of individuals for the purpose of identifying the drivers of variation in the
57 human immune system.

58 **Human Immune Variation**

59 The human immune system has a high degree of variation present between
60 individuals. This variation manifests at the cellular and intra-cellular level, such as the
61 differences in the relative frequency of different leukocyte populations and subsets,
62 variation in the transcriptional and translational profiles of leukocyte subsets and
63 variation in the functional capacity and polarization in response to immunological
64 challenges such as vaccines [9]. This variability results in individuals with a different
65 intrinsic susceptibility to different diseases. As an example, the concept of individuals
66 being “Th1-biased” or “Th2-biased” is an articulation of this intrinsic variability, albeit
67 a simplification of the multiple potential states the human immune system could
68 adopt. More broadly, we consider the constellation of multiple cellular and molecular
69 immune parameters within an individual to constitute the immune phenotype of that
70 individual.

71 The use of high detail longitudinal analysis of the human immune system has
72 revealed this immune phenotype to be highly stable. In principle, the variation
73 present in immune phenotypes within the human population could be generated
74 through either of two mechanisms. First, the human immune system could undergo a
75 continual flux within each individual (producing high levels of intra-individual
76 variation), a process which would in turn produce a high degree of inter-individual
77 variation (**Figure 1a, Online Video 1**). Alternatively, the immune profile of individuals
78 could have relatively low levels of temporal variation in a given individual, but with
79 highly divergent phenotypes between individuals (**Figure 1b, Online Video 2**). In this
80 case, an individual immune phenotype would constitute a ‘stability island’, around
81 which the immune parameters remain fixed over time. Multiple immune phenotyping
82 studies, investigating the relative frequency of leukocyte subsets and activation
83 states, have found that while inter-individual variation is high, the immune profile of

84 each individual is remarkable stable, even over the course of multiple months [10-13],
85 favoring the second mechanism described above.

86 The long-term stability of an individual's immune system in the absence of
87 immunological challenge does not require that stability is maintained during infection.
88 One way to experimentally determine how an acute challenge impacts the immune
89 system in humans is to use vaccination. This enables the global assessment of the
90 immune system prior to, during, and following a known immunological stimulus [14].
91 There is a general pattern that emerges from these systems vaccinology studies that
92 demonstrates the elastic response of the human immune system to acute challenge.
93 The early phase of the immune response (0-3 days post vaccination) is characterised
94 by the expansion of circulating antigen-presenting innate immune cells, which
95 activate T cells. The resultant expansion of T cells and the production of antibody-
96 secreting plasma cells peak at 6-10 days post-vaccination, and are followed by a
97 return to the pre-vaccination baseline [12, 13, 16-20]. The long-term functional
98 memory T and B cells that persist [21, 22], are not present at a sufficient numbers to
99 alter the cellular composition of the circulating immune system [12, 13, 16, 23-25]. In
100 principle, an individual's immune profile could elastically return to the original stable
101 state following this immunological challenge (**Figure 1c, Online Video 3**), or the
102 transient changes caused during infection could cause it to settle into a new stable
103 immunological state (**Figure 1d, Online Video 4**). The data for both gastrointestinal
104 infection (traveller's diarrhoea) and influenza vaccination in people suggest that the
105 'elastic rebound' model is appropriate [13]. In other words, individuals will generally
106 return to an immune phenotype similar to their original state after an acute immune
107 stimulus, although chronic infections may be an exception (see dedicated section
108 below). The presence of these two characteristics, longitudinal stability and elastic
109 rebound after challenge, allows the further dissection of which factors shape
110 diversity.

111 **Intrinsic Factors Influencing the Human Immune System**

112 *Age*

113 With advancing age the function of the immune system declines, rendering older
114 persons more susceptible to infection and less able to generate protective immunity
115 after vaccination [26]. Population immunology studies that have included age as a
116 variable have found that aging impacts on ~20% of immunological traits, and overall

117 contributes ~5% of total immune variation [10, 11, 13, 27]. Analysis of the specific
118 traits impacted demonstrates that the ageing human immune system is largely
119 characterized by a decrease in the frequency of precursor cells, and an increase in
120 the number of T cells with an activated or memory phenotype [10, 11, 13, 27, 28].
121 The changing immune landscape over time is likely driven by a number of biological
122 mechanisms. As the immune system is constantly interacting with both pathogens
123 and commensals, it is thought that these interactions play a key role in shaping the
124 human immune system, such as by increasing the number of memory cells with age
125 [29]. However, work with germ-free mice demonstrates that T cells that have not
126 'seen' their cognate antigen can acquire a memory phenotype via homeostatic
127 expansion [30, 31]. Memory phenotype cells are found in human cord blood
128 suggestive of homeostatic expansion [32], but the assumed sterility of the *in utero*
129 environment is questionable [33]. Ageing is characterized by an increase in
130 proinflammatory cytokines, which alter the environment of the immune system. This
131 in turn may alter the phenotype and function of immune cells in an antigen-
132 independent way [34]. Because humans maintain their naïve T cell pool for decades
133 by peripheral T cell division alone [35], and as the long-term impact of a single
134 immune challenge is minute in the immune landscapes of healthy people [12, 13], it
135 is plausible that the age-dependent skewing of the immune system from a naïve to
136 memory phenotype may occur, in part, without recognition of cognate antigen or as a
137 result of cross-reactivity [36, 37].

138 Despite the clear role for non-genetic factors shaping the ageing immune system,
139 cellular phenotypes associated with ageing, for example the decline in thymic output
140 and thus numbers of recent thymic emigrants, have high heritability [10, 11, 13, 38].
141 This suggests that the interplay of genetic and non-genetic factors is important for the
142 development of ageing phenotypes. While the role for selection of genes that impact
143 ageing phenotypes is an area of intense debate [39], human longevity studies show a
144 clear role of genetics in impacting how well we age [40]. Specific studies on how
145 genetics shape the human immune system have yet to be realized. Nevertheless,
146 single nucleotide polymorphisms associated with longevity are also linked with
147 functional changes in the immune system [40-42], suggesting a link between
148 immunity, inflammation and health in older age.

149 Sex

150 It is clear that there is sexual dimorphism in immune-related pathology. In general,
151 autoimmune diseases are more common in women than men [43], while men are
152 more susceptible to infections caused by viruses, bacteria, parasites and fungi [44].
153 There is also evidence that women produce higher antibody titers in response to
154 vaccination than men, suggesting sex can impact immune function [44]. Despite clear
155 differences in functional outcome, the differences in the cellular composition of the
156 immune system between men and women are few; women tend to have a higher
157 CD4⁺ T cell count, and lower natural killer (NK) cell counts than men, but otherwise
158 no consistent differences are observed [13, 27, 45], suggesting sex has a larger
159 influence on disease outcome than immune composition. This is supported by large
160 scale phenotyping of clinical and immunological parameters in parallel, that indicates
161 that sex has the largest effect on clinical phenotypes [27]. Interestingly, the functional
162 and cellular differences between the sexes are more pronounced in men with high
163 testosterone [46], and disappear after menopause [44, 45], suggesting that the sex
164 hormones are responsible for these changes, although precisely how this is mediated
165 has yet to be elucidated.

166 **Genetic Control Over Human Immune Variation**

167 A likely driver for individual-to-individual variation in the settings of the immune
168 system is genetic variation. Indeed, major studies into the genetic basis of variation in
169 the cellular and molecular composition of the human immune system have found that
170 genetic variation (e.g. common and rare variants, variations in copy number)
171 accounts for 20-40% [10, 11, 47, 48] of total immunological variation (**Figure 2**).
172 These studies have used either a twin or familial structure to estimate heritability. The
173 twin studies use the classical ACE model (see **Glossary**), while Orrù *et al.* use a
174 kinship matrix based on genotype to define relatedness [11]. Heritability estimates
175 should therefore be considered with a broad definition of genetic variation, including
176 (but not limited to) common, copy number and rare variants; heritable epigenetic
177 modification and genotype-genotype interactions (for example, interactions between
178 Killer immunoglobulin-like receptor gene variants and HLA variants determine the
179 severity of CMV infection [49]).

180 The range of heritability estimates, from 20 to 40% of total immunological variance, is
181 likely to reflect both biological differences in the populations being measured and
182 methodological differences in sampling and the choice of which immunological

183 parameters to include in the study of overall immune diversity. In this regard, it is
184 notable that the highest estimates came from analysis of the Sardinian population
185 (which is a genetic isolate). The lowest level of heritability was reported by Ye *et al.*
186 [47]. There were several features unique to this study: (i) transcriptome-level
187 analysis; (ii) tightly controlled sampling protocol (with only sampling between the
188 hours of 07:30-08:30, from fasted participants); (iii) cells were examined both *ex vivo*
189 and after stimulation and (iv) heritability was estimated from principal components
190 based on ancestry. It is tempting, when comparing the lower heritability in Ye *et al.* to
191 the other studies, to speculate that the transcriptome might well have a larger total
192 variance than the cellular composition of the immune system. For instance, it seems
193 plausible that the absolute expression of *CD4* mRNA may vary more markedly than
194 the frequency of CD4⁺ T cells. In support of this proposition, an older study using only
195 basic immune populations found a much higher heritability estimate of ~70% [50].
196 This would indicate a hierarchy of genetic influence, with the strongest effects being
197 felt at the leukocyte population level, moderate effects at the level of subsets and
198 activation potential, and the weakest effects at the level of transcription. By inference,
199 environmental effects are likely to be strongest at the molecular level and would have
200 the weakest potential to influence major leukocyte populations. An exception may be
201 cis-acting regulatory polymorphisms, with some of the highest heritability found for
202 single protein expression markers and their corresponding gene (for example CD39
203 expression with *ENTPD1*, the CD39 gene, variants [38]). While it is debatable as to
204 which level of immune variation is the most relevant, given that the basic unit of
205 immunity is the cell it seems reasonable to us to focus on the variation present at the
206 level of cellular subset and function, rather than lower-order (transcriptome, single
207 gene expression) or higher-order (basic leukocytes) variation.

208 Beyond gross heritability analysis, several studies have explored the relationship
209 between specific genetic variants and individual immune parameters. To date, only
210 associations with common single nucleotide polymorphisms (SNP) have been
211 explored [11, 38, 47]. The relatively low number of individuals assessed in this
212 manner (compared to disease-based genome-wide association studies), and the
213 multiple classes of genetic variants missing from these studies (**Box 3**), preclude final
214 conclusions being made – there is more to be found from genetic analyses. However,
215 as may be expected, many of the SNPs identified that are associated with changes in
216 immune cell types are in or near genes with known functions in the immune system,

217 and in particular within genes that have been previously associated with
218 autoimmunity, inflammatory diseases and susceptibility to infection [11, 38]. These
219 findings strongly support the supposition that diversity in the immune system
220 contributes to the relative risk of the individual to develop immunological disease.

221 **Environmental Influences on Human Immune Variation**

222 The estimate of the contribution heritable (20-40%) and intrinsic factors (~5%) make
223 to the observed immunological variation suggests a major role for environmental
224 factors. Unlike genetic and intrinsic factors, which are readily identifiable and
225 measureable, dissecting out the individual environmental factors that may play a role
226 is confounded by the limitless list of putative environmental factors that could be
227 considered, the difficulty in collecting data for any of these factors, and the unknown
228 timeframe over which environmental factors may be acting. Nevertheless, there are
229 indications of the potency and nature of environmental contributors to immunological
230 variation. We recently assessed the immunological variation that is present between
231 opposite sex couples living with a child and found that the degree of variation was
232 50% lower than is observed in the general public [13], indicating that a convergence
233 in immune status occurs during cohabitation (**Figure 3** and **online video 5**). Even
234 with the proviso that couples are not randomly assorting, this data indicates that
235 environmental factors are potent, that they can modify the adult immune system, and
236 that they can be observed at the level of the household. The cohabitation effect is
237 likely to be driven through the accumulated impact of multiple smaller factors, as the
238 reduction of diversity evident within each couple was unique [13] (while convergence
239 driven by a single environmental factor would be expected to show a consistent effect
240 across multiple couples).

241 There are multiple promising candidates for the household environmental factors
242 capable of shaping the variation present in the human immune system. One of the
243 strongest is likely to be chronic viral infections, with discordance in cytomegalovirus
244 infection a major driver of immune diversity between monozygotic twins [10] (see
245 detailed section below). Other factors that are shared within household members and
246 are known to alter the immune state, and which could therefore drive household-level
247 immunological convergence, include smoking status [51], well-being [52, 53], pet
248 ownership [54], physical activity [55], obesity [13], diet [56] and environmental
249 pollutants [57, 58]. Each of these factors is also likely to influence the composition of

250 that individual's microbiota. The indirect microbiota convergence between individuals
251 that share an environment (and thus the same microbiota modifiers) is also
252 accompanied by direct transmission of microbiota between individuals in close
253 proximity, enhancing microbiota convergence [59-62]. With the growing evidence that
254 the microbiota composition, especially within the gut, influences the immune system
255 [63-67], it is plausible that microbiota changes act as an integrator of multiple
256 environmental factors, in turn shaping diversity within the human immune system.
257 Seasonal change is another environmental factor that can influence immune variation
258 [68], however it is unlikely to be captured in most study designs, as it would not
259 induce variation between individuals living in the same region at a given time-point.

260 Bacteria and viruses are both able to establish chronic infections. For example
261 *Mycobacterium tuberculosis* and cytomegalovirus (CMV) are able to persist long-term
262 in humans. Such infections can be established when the pathogen is able to evade
263 sterilizing immunity, followed by formation of an equilibrium between the pathogen
264 and host in which the inflammatory response is reduced to a level that can control the
265 pathogen without causing excess damage to the host [69, 70]. The balance between
266 host and chronic infectious agents is a complex one; ultimately the pathogen requires
267 a live host for its continued survival so must exist in a balance of not triggering overt
268 immunity and clearance, and not overcoming the host with disease. This constant
269 interaction with the host immune system has an indelible impact on the immune
270 landscape.

271 There are a number of chronic viral infections that rarely cause disease in healthy
272 hosts with symptoms only becoming apparent in immunocompromised individuals
273 [69]. Despite the lack of clinical disease, this long-term dynamic interaction between
274 host and virus has significant impact on altering the cellular composition of the
275 immune system. The best characterised of these persistent infections are CMV and
276 Epstein-Barr virus (EBV). CMV influences the immune system directly, by increasing
277 the frequency of CMV-specific CD4⁺ and CD8⁺ T cells to around 5% of the total T cell
278 pool in people under 55 years old [71], the frequency of CMV specific cells increases
279 further with age suggesting a dynamic host-pathogen interaction that continues to
280 shape the immune system [29]. This effect is likely due to the combined impact of
281 thymic involution with the continued stimulation of CMV-specific T cells clones
282 throughout life [72]. The effects of CMV are not limited to antigen-specific T cells;
283 global analysis of the immune systems either unrelated individuals or monozygotic

284 twins that are discordant for CMV demonstrate that presence of this virus alters the
285 level of certain serum cytokines, such as IL-6, and frequency of many cell types
286 including $\gamma\delta$ T cells, granulocytes and memory CD8⁺ T cells [10, 27]. It is clear that
287 CMV has significant effects on the composition of the immune system, and intriguing
288 that it seems to be the most extreme example of a persistent infection impacting the
289 host in this way. A caveat for the role of CMV in these studies, is that their
290 concordance for other chronic infections was not assessed [10]. However, individuals
291 infected with herpes simplex virus do not exhibit the type of changes in their
292 circulating T cell compartment that are driven by CMV [73]. Persistent EBV infections
293 that have not caused acute disease (infectious mononucleosis) also result in the
294 accumulation of virus-specific CD4⁺ and CD8⁺ T cells, but not as high a frequency as
295 CMV specific T cells [74, 75].

296 **Concluding Remarks**

297 The advent of large-scale high-depth immune profiling (“population immunology”) has
298 allowed a first insight into the scope and nature of immune diversity within the human
299 population. In the case of systems vaccinology, this variation has a clear potential
300 impact on health, with altered efficacy of vaccination against infectious diseases [12,
301 18]. However, variation in the human immune system is likely to not only influence
302 responses to infections, but also to impact on susceptibility to autoimmunity, allergy,
303 cancer and the inflammatory diseases of ageing, such as diabetes, cardiovascular
304 disease and neurodegeneration. To a certain degree this variation, and the resulting
305 disease susceptibility, is ‘locked in’, owing to the influence of genetic and intrinsic
306 factors. However, a surprisingly high influence of environmental control has also
307 been observed, such as is revealed by the effect of cohabitation [13]. New studies
308 need to be performed to identify the distinct interactions between environment and
309 immune variation, however for parameters such as microbiome [64], diet [56], air
310 pollution [57] and even anxiety [53], it is likely that much of the influence they have
311 over disease susceptibility will lie in their capacity to modulate the immune system.
312 Dissections of the particular environment-immune interactions that constitute disease
313 risk have the potential to allow targeted modification of the immune system by
314 lifestyle and environmental modulation (see Outstanding Questions). Regulation of
315 environmental influences over the immune system through approaches such as
316 dietary modification, restricting exposure to certain pollutants and monitoring
317 microbiota changes would combine aspects of personalized medicine with

318 preventative medicine, and could have a profound impact on the way immunological
319 health is managed.

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511

512 **Figure legends**

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516

517 **Box 1. Limitations in SNP Genotyping Platforms**

518 The SNP genotyping platforms typically used for genome-wide association studies
519 place an emphasis on common single nucleotide (>5% allelic frequency), which
520 accounts for $\pm 90\%$ of genetic variation in the human population [76]. This excludes
521 certain classes of genetic variation, notably copy number variation and rare variants,
522 which, while less common, may exert a disproportionate degree of influence over the
523 immune system. SNP genotyping platforms can also struggle with immunological loci,
524 owing to the enrichment for gene duplication events [77, 78]. For example, the
525 rs1050501 SNP in *FCG2RB*, confers susceptibility to systemic lupus erythematosus
526 and protection from malaria [79]. The pseudogene *FCGR2C* has an identical
527 nucleotide sequence in this exon, but is homozygous for the ancestral allele.
528 Genome-wide SNP platforms will discard rs1050501 at the quality control stages, as
529 it will not follow Hardy-Weinberg equilibrium (that allele frequencies will remain
530 constant across generations in the absence of selection pressure) due to
531 contaminating signal from *FCGR2C*. These effects place a technical burden on
532 estimating the contribution of such loci, which may cause an under-estimation of the
533 genetic contribution of immunological loci.

534 **Figure 1. Opposing Models for the Properties of Human Immune Variation.**

535 Variation in the immune system can be represented as different points that can be
536 occupied on an 'immunological landscape'. **A.** The stochastic model is based on a
537 fluctuating immune system, where apparent diversity is observed due to taking single
538 snap-shots of an immune system in flux. **B.** The stability model is based on
539 individual-to-individual variation reflecting a consistent longitudinal difference, with
540 only minor changes occurring within healthy individuals monitored over time. The
541 stability model raises additional possibilities of elastic or inelastic stability. **C.** The
542 elastic stability model is based on individuals maintaining distinct immune variation,

543 consistently across time. While immunological disturbances (such as infection) can
544 transiently alter the state, following the end of the disturbance, individuals 'rebound'
545 to the original state. **D.** The inelastic stability model also has maintenance of distinct
546 immune variation over time, however immunological challenges can disrupt the
547 default state, resulting in a different stable state being achieved. The white flashes in
548 C and D represent a immunological challenge, such as vaccination, or infection.

549 **Figure 2. Heritability of Immune Traits Across Different Populations and Study**
550 **Designs.** Published heritability estimates for individual immune traits. For each study,
551 the study design, numbers of individuals and measurement strategy (blue, twin study;
552 orange, family study) is displayed. For Ye *et al.*, heritability estimates for each
553 individual trait were unavailable; the 'global' reported estimate is shown; this study
554 used a population study design. Boxes depict interquartile range (IQR) and median.
555 The whiskers extend up to the furthest datapoint or 1.5 x IQR, in which case outliers
556 are plotted with dots. Abbreviations used: MZ monozygotic; DZ dizygotic.

558

559 **Figure 3. Interaction of Genetics, Age and Environment on Immune Variation.**
560 Immune variation can be modeled as an 'immunological landscape', with individuals
561 occupying different points on the landscape. A synthesis of current data would place
562 the role of genetics as influencing the initial starting point on the immune landscape.
563 This initial variation is molded by age, with predictable changes occurring over time,
564 driving the same changes in different individuals. Local environment strongly modifies
565 other variables, leading to factors such as cohabitation (pink) resulting in
566 convergence between different individuals to a closer point on the immunological
567 landscape.

568 **Online Video 1. The Stochastic Model of Human Immune Variation.** Variation in
569 the immune system can be represented as different points that can be occupied on
570 an "immunological landscape". The stochastic model is based on a fluctuating
571 immune system, where apparent diversity is observed due to taking single 'snap-
572 shots' of an immune system in flux. Under this model, the relative diversity between
573 individuals will change dramatically over time, as each individual constantly moves
574 through the immunological landscape.

575 **Online Video 2. The Stability Model of Human Immune Variation.** Variation in the
576 immune system can be represented as different points that can be occupied on an
577 “immunological landscape”. The stability model is based on individual-to-individual
578 variation reflecting a consistent longitudinal difference, with only minor changes
579 occurring within healthy individuals monitored over time. In this model, the immune
580 system of individuals can vary with time, but this variation is largely contained to a
581 limited region on the immunological landscape (visualized through a depression in
582 the landscape). The current data obtained from population immunology studies
583 supports this model over the one presented in Video 1.

584 **Online Video 3. The Elastic Stability Model of Human Immune Variation.**
585 Variation in the immune system can be represented as different points that can be
586 occupied on an “immunological landscape”. The stability model is based on
587 individual-to-individual variation reflecting a consistent longitudinal difference, with
588 only minor changes occurring within healthy individuals monitored over time. In this
589 model, the immune system of individuals can vary with time, but this variation is
590 largely contained to a limited region on the immunological landscape during the
591 homeostatic context (visualized through a depression in the landscape). Despite this
592 stability during homeostasis, the immune system can radically change during
593 infection (visualized through the individuals turning black). Under the elastic stability
594 model, these immunological disturbances are only transient, and following the end of
595 the disturbance, individuals “rebound” to the original homeostatic state. Current data
596 seems to support this model in most cases.

597 **Online Video 4. The Inelastic Stability Model of Human Immune Variation.**
598 Variation in the immune system can be represented as different points that can be
599 occupied on an “immunological landscape”. The stability model is based on
600 individual-to-individual variation reflecting a consistent longitudinal difference, with
601 only minor changes occurring within healthy individuals monitored over time. In this
602 model, the immune system of individuals can vary with time, but this variation is
603 largely contained to a limited region on the immunological landscape during the
604 homeostatic context (visualized through a depression in the landscape). Despite this
605 stability during homeostasis, the immune system can radically change during
606 infection (visualized through the individuals turning black). Under the inelastic stability
607 model, these immunological disturbances create a lasting imprint on the individual’s

608 immune variation, with a new homeostatic state being established at the resolution of
609 infection. Infections that can establish chronicity are a good example of this model.

610 **Online Video 5. Interaction of Genetics, Age and Environment on Immune**
611 **Variation.** Immune variation can be modeled as an “immunological landscape”, with
612 individuals occupying different points on the landscape. A synthesis of current data
613 would place the role of genetics as influencing the initial homeostatic stability region
614 on the immune landscape. Individuals show minor variation within their unique
615 stability region over the short-term, however over the long-term the region of stability
616 is molded by age, with predictable changes occurring over time, driving the same
617 changes in different individuals (visualized by the downwards movement of the
618 depression). In addition to the effect of age, local environment strongly modifies other
619 variables, with factors such as cohabitation (with the pink flash indicating the initiation
620 of cohabitation between the visualized individuals) resulting in convergence between
621 the individuals to a closer point on the immunological landscape.

622

623 **Glossary**

624 **Population immunology** : The study of human immunology where the primary focus
625 is to understand the diversity of immune states and responses across a large population.
626 Population immunology studies need to combine a high depth (e.g., detailed analysis
627 of leukocyte population subsets or function) with large population size (typically
628 between 100 and 1000 individuals) in order to identify the degree of immune variation
629 and potential drivers.

630 **ACE model:** In the ACE model, total phenotypic variance is considered to be the
631 sum of the variances attributed to genetic (A), common environmental (C) and unique
632 environmental (E) factors. Twins share C and E, with monozygotic twins fully sharing
633 A and dizygotic twins sharing 50%. Thus the assessment of multiple sets of
634 monozygotic and dizygotic twins allows an estimation of the proportion of total
635 phenotypic variance that is driven by genetic variation.

636