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Commercially Available Outbred Mice for Genome-Wide Association Studies

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Abstract

Genome-wide association studies using commercially available outbred mice can detect genes involved in phenotypes of biomedical interest. Useful populations need high-frequency alleles to ensure high power to detect quantitative trait loci (QTLs), low linkage disequilibrium between markers to obtain accurate mapping resolution, and an absence of population structure to prevent false positive associations. We surveyed 66 colonies for inbreeding, genetic diversity, and linkage disequilibrium, and we demonstrate that some have haplotype blocks of less than 100 Kb, enabling gene-level mapping resolution. The same alleles contribute to variation in different colonies, so that when mapping progress stalls in one, another can be used in its stead. Colonies are genetically diverse: 45% of the total genetic variation is attributable to differences between colonies. However, quantitative differences in allele frequencies, rather than the existence of private alleles, are responsible for these population differences. The colonies derive from a limited pool of ancestral haplotypes resembling those found in inbred strains: over 95% of sequence variants segregating in outbred populations are found in inbred strains. Consequently it is possible to impute the sequence of any mouse from a dense SNP map combined with inbred strain sequence data, which opens up the possibility of cataloguing and testing all variants for association, a situation that has so far eluded studies in completely outbred populations. We demonstrate the colonies’ potential by identifying a deletion in the promoter of H2-Ea as the molecular change that strongly contributes to setting the ratio of CD4+ and CD8+ lymphocytes.

Introduction

The design of an ideal population for gene mapping involves balancing the avoidance of rare alleles with the requirement for rapid linkage disequilibrium (LD) decay. High rates of LD decay are found in populations with large effective population sizes and many generations of random mating that accumulate recombinants to break up correlations between genotypes. Unfortunately, a necessary corollary is the presence of rare alleles as allele frequencies drift to extremes and new, rare, alleles arise as a result of population structure, caused either by recent admixture or uneven population density. The more rare alleles in a population, and the more they contribute to phenotypic variation, the more difficult it will be to detect quantitative trait loci (QTLs) using genome-wide association strategies that genotype only common alleles [1].

The best strategy might seem to be to choose animals from highly divergent populations, such as wild mice caught in many locations [2], or from inbred lines derived from highly genetically divergent progenitor strains. This maximizes genetic diversity and seeks to overcome the limitations of using only a subset of the variation present in wild populations. However, mice from different populations will have a high proportion of private variants present in one population only. LD decay for the latter private variants will depend solely on recombinants accumulated during the creation of the colony, while LD decay for the former, common, variants is boosted by the ancestry of the founding populations. Furthermore, the power to detect a genetic effect increases with the minor allele frequency (MAF) of the causal variant. It follows that high power and mapping resolution is best obtained by using animals from the same mating population to reduce the number of private alleles. A related phenomenon is population structure, caused either by recent admixture or uneven degrees of relatedness, both of which should be avoided.

Commercial mouse breeders maintain large colonies of outbred mice that may have the necessary genetic structure. LD in some outbred stocks has been shown to allow high-resolution mapping [3], sufficient to identify genes [4]. Importantly, most outbred stocks are known to derive from animals from a single population, such as the ‘Swiss’ stocks which descend from two males and seven females imported from Lausanne, Switzerland [3], indicating that
the proportion of private alleles may be low. Figure 1 summarizes the known relationship between commercially available outbred stocks as of 2007 (the time of this study) and additional information is given in Text S1.

## Results

### Stocks, colonies, and genetic markers

Table 1 lists the populations that we obtained for this study and the numbers of animals we used. We included three control colonies.

### Origins of outbred mice

**Commercially Available Mice**

| Outbred stocks | NMRI | CFW and MF1 | ICR and CD1 | NIHs | ND4 | SW | Non-Swiss (TO, SABRA, CF1, NISA, OF1) |
|----------------|------|-------------|-------------|------|-----|----|-------------------------------------|

### Abbreviations

- Awi: Ace Animals, Inc.
- Bi: BiB Universal, Inc.
- CAR: Centre Anticancerque Romand
- CD1: Cederon Derived 1
- CF: Canworth Farms
- CFW: Canworth Farms White
- CSBL: Central Institute for Laboratory Breeding
- Crl: Charles River Laboratories
- CSAL: Centre de Service des Animaux de Laboratoire
- DE: Germany
- DK: Denmark
- Fa: Hilltop Lab Animals, Inc.
- Harlan: Harlan Sprague-Dawley, Inc.
- I.C.I.: Imperial Chemical Industries Americas
- ICR: Institute of Cancer Research (Ken Chesley Cancer Centre)
- MA: Møllerup and Bornholm
- MF1: Mäuse-Forschung
- ND: Nude
- NIH: National Institutes of Health
- NMRI: National Medical Research Institute
- NSA: Non-Swiss Albino
- NY: New York
- OF1: OXOAN, OF1
- OX: Oxford Laboratory Animal Colonies
- Rcc: Research and Consulting Company, Ltd.
- RF: Rockland Farms
- RJ: Centre d’Etude de l’Animal
- RPMI: Roswell Park Memorial Institute
- Sci: Scandinavia
- Tac: Tac份for, Inc.
- TO: Tuck Ordinary
- US: United States
- WI: Wisconsin Institute
- Win: Winslow Veterinary Research GmbH & Co

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**Figure 1. Ancestry of commercially available outbred stocks.** Most outbreds have a common origin: they descend from a single Swiss colony of 200 mice from which 2 males and 7 females were imported to the Rockefeller Institute for Medical Research in New York [5]. Outbred Swiss stocks currently available include NMRI, CFW, MF1, CD1, ICR, NIHs, ND4 and SW. Not all outbreds descend from Lausanne in Switzerland. Non-Swiss strains include CF-1, NISA, OF1, SABRA and TO. Details on the origin of colonies are provided in Text S1.

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However, without systematic characterization of the genetic architecture of commercially available outbred mice it is not possible to evaluate the suitability of any particular colony for genome-wide association. In this paper we evaluate 66 populations to determine (i) whether inbreeding and population structure preclude the use of the population; (ii) whether linkage disequilibrium (LD) enables high-resolution mapping; (iii) whether the proportions of common and rare variants are favorable for genome-wide association mapping. In order to assess the latter, we tested the hypothesis that the outbred colonies are descended from a common source: the laboratory inbred strains. Populations in which this assumption holds true, and which have low levels of LD, are the most suitable for high-resolution mapping. Finally, we show how commercially available outbred mice can be used to go from genetic association to molecular lesion by identifying a deletion in the promoter of *H2-Eα* as the molecular change that strongly contributes to setting the ratio of CD4+ and CD8+ lymphocytes.

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**Author Summary**

We show that commercially available mice are a resource for detecting single genes by genome-wide association. We surveyed 66 populations and identified those with properties conducive to high-resolution mapping. Importantly, we show that the same alleles contribute to variation in different colonies, so that when mapping progress stalls in one colony, another can be used in its stead. As a proof of principle, we detect the same QTL in different colonies influencing CD4+/CD8+ ratios and refine this mapping to the gene level. We show that a deletion in the promoter of *H2-Eα* is the molecular change that strongly contributes to setting the ratio of CD4+ and CD8+ lymphocytes. Our results make it possible for geneticists to make informed choices on the use of colonies for genome-wide association studies of complex traits in mice.
| Colony                     | No. | Mk. | M/F | Breeding   | Gps. | Size | Date     | Mean | MAF | LD decay radius | Het. | Pct fail HWE | In-breeding coef | Struct. | Status 12/09 |
|----------------------------|-----|-----|-----|------------|------|------|----------|------|-----|----------------|------|--------------|-----------------|---------|-------------|
| Aai:ICR-US                 | 24  | 351 | 1/2 | Circular   | 10   | 1200 | 24/10/2007 | 0.026 | 1.88 | 0.08           | 2.27 | 2.76         |                 |         |             |
| BKW-UK                     | 48  | 351 | 1/6 | Poiley     | 3    | 925  | 12/10/2007 | 0.024 | 1.12 | 0.12           | 2.27 | 8.78         |                 |         |             |
| BomTacNMRI-DK_151          | 23  | 351 | 1/3 | Poiley     | 6    | 453  | 24/10/2007 | 0.068 | 1.07 | 0.16           | 1.70 | –5.68        |                 |         |             |
| BomTacNMRI-DK_160          | 24  | 351 | 1/3 | Poiley     | 6    | 1038 | 28/09/2007 | 0.075 | 0.87 | 0.15           | 1.98 | 4.57         |                 |         |             |
| CrlHe:CD1-IL               | 20  | 351 | 1/1 | Poiley     | 4    | 16   | 27/11/2007 | 0.008 | 2.78 | 0.01           | 0.57 | –16.50       |                 |         |             |
| Collaborative Cross        | 109 | 351 |      |            |      |      | 07/11/2008 | 0.254 | NA  | 0.19           | 89.24 | 67.28        |                 |         |             |
| CrICD1(UCR)-UK             | 48  | 7/25 | 1/6 | Poiley     | 3    | 925  | 10/08/2009 | 0.024 | 0.57 | 0.15           | 3.97 | 13.73        |                 |         |             |
| CrICD1(UCR)-DE             | 48  | 351 | 1/3 | Poiley     | 6    | 453  | 24/10/2007 | 0.068 | 1.12 | 0.04           | 2.27 | 8.78         |                 |         |             |
| Crl:CD1(ICR)-UK            | 48  | 351 | 1/1 | IGS        | 1    | 1950 | 01/08/2007 | 0.126 | 1.00 | 0.27           | 3.97 | 4.40         |                 |         |             |
| Crl:CD1(ICR)-US_iso        | 30  | 351 | 1/3 | Poiley     | 6    | 36   | 10/08/2009 | 0.075 | 0.87 | 0.15           | 2.27 | 4.57         |                 |         |             |
| Crl:CF1-UK                 | 48  | 4/15 | 705 | 22/11/2007 | 0.194 | 2.37 | 0.35 | 6.80 | 10.04         |       |             |                 |         |             |
| Crl:CFW(SW)-US_P08         | 48  | 351 | 1/6 | Poiley     | 6    | 36   | 10/08/2009 | 0.075 | 0.87 | 0.15           | 2.27 | 4.57         |                 |         |             |
| Crl:CD1(UCR)-R16           | 24  | 351 | 1/3 | Poiley     | 6    | 453  | 24/10/2007 | 0.068 | 1.12 | 0.04           | 2.27 | 8.78         |                 |         |             |
| Crlj:CD1(ICR)-JP           | 48  | 351 | 02/12/2008 | 0.073 | 1.34 | 0.21 | 7.08 | 4.61 |       |       |             |                 |         |             |
| HanRcc:NMRI-CH             | 48  | 351 | 1/3 | Poiley     | 6    | 453  | 24/10/2007 | 0.068 | 1.12 | 0.04           | 2.27 | 4.57         |                 |         |             |
| Heterogeneous Stock        | 12  | 351 |      |            |      |      | 07/11/2008 | 0.254 | NA  | 0.19           | 89.24 | 67.28        |                 |         |             |
| Hla:ICR(CD-1)-DE           | 48  | 351 | 1/3 | Random     | 1    | 2500 | 26/10/2007 | 0.098 | 0.79 | 0.21           | 4.82 | 3.13         |                 |         |             |
| HsICD1(CD-1)-E              | 53  | 351 | 1/2 | Random     | 1    | 750  | 03/11/2008 | 0.153 | 1.08 | 0.29           | 5.10 | 2.13         |                 |         |             |
| HsICD1(CD-1)-FR             | 48  | 351 | 1/3 | Random     | 1    | 1563 | 14/11/2007 | 0.147 | 1.49 | 0.26           | 5.38 | 3.49         |                 |         |             |
| HsICD1(CD-1)-FR             | 64  | 351 | 1/2 | Random     | 1    | 2000 | 06/08/2007 | 0.155 | 0.99 | 0.28           | 5.38 | 3.60         |                 |         |             |
| HsICD1(CD-1)-FR             | 20  | 600K|      |            |      |      | 0.105           | 0.22 | 1.53 | –1.82         |       |             |                 |         |             |
| HsICD1(CD-1)-IL             | 48  | 351 | 1/3 | Random     | 1    | 500  | 27/11/2007 | 0.143 | 1.34 | 0.29           | 3.68 | –6.55        |                 |         |             |
| HsICD1(CD-1)-IT             | 48  | 351 | 1/3 | Random     | 1    | 1450 | 16/11/2007 | 0.162 | 1.07 | 0.28           | 4.82 | 7.52         |                 |         |             |
| HsICD1(CD-1)-MX             | 48  | 351 | 1/3 | Random     | 1    | 500  | 05/12/2008 | 0.147 | 1.24 | 0.28           | 3.97 | –0.34        |                 |         |             |
| HsICD1(CD-1)-US             | 48  | 351 | 1/2 | Random     | 1    | 2592 | 04/08/2008 | 0.149 | 1.05 | 0.28           | 5.38 | 3.43         |                 |         |             |
| HsdND4-US                   | 48  | 351 | 1/3 | Random     | 1    | 1002 | 04/08/2008 | 0.036 | 1.79 | 0.07           | 2.27 | 4.89         |                 |         |             |
| HsdNIHS-UK_C               | 15  | 351 | 1/3 | Random     | 1    | 66   | 05/12/2008 | 0.055 | 1.02 | 0.11           | 1.70 | 6.36         |                 |         |             |
| HsdNIHS-UK_G               | 33  | 351 | 1/3 | Random     | 1    | 33   | 05/12/2008 | 0.084 | 2.04 | 0.11           | 3.12 | –5.09        |                 |         |             |
| HsdNIHS-US                  | 48  | 351 | 1/2 | Random     | 1    | 1163 | 04/08/2008 | 0.011 | 2.45 | 0.19           | 9.92 | –18.01       |                 |         |             |
| HsdNIHSC-IL                 | 12  | 351 | 1/2 | Poiley     | 4    | 16   | 27/11/2007 | 0.047 | 1.05 | 0.02           | 0.57 | 3.11         |                 |         |             |
populations, with known genetic characteristics: 12 Heterogeneous Stock (HS) mice [6], 109 Collaborative Cross (CC) mice [7], 94 inbred strains [8] and a population of wild mice caught from multiple sites in Arizona that is likely to represent a fully outbred colony. Some colonies have been culled since the sampling date and these are indicated by an * in the column headed Status 12/09. We provide four genetic measures for each colony: the mean minor allele frequency (Mean MAF), heterozygosity (Het) and an inbreeding coefficient. The LD decay radius, a measure of the colony's suitability for high-resolution mapping, is the average physical separation between SNPs beyond which the squared correlation coefficient drops below 0.5. Colonies where we found evidence for population structure are show in the column Struct. The multi-dimensional scaling of IBS pairwise distance matrices on which this is based is shown in Figure S1.

For each colony listed in column one, we report the number of animals we used (No.); the number of markers analysed (genome-wide marker sets were used for six populations); the sex ratio (Male to Female); the supplier's breeding scheme (where available); the number of groups of animals used in that scheme and the size of the colony. Some colonies have been culled since the sampling date and these are indicated by a * in the column headed Status 12/09. We provide four genetic measures for each colony: the mean minor allele frequency (Mean MAF), heterozygosity (Het) and an inbreeding coefficient. The LD decay radius, a measure of the colony’s suitability for high-resolution mapping, is the average physical separation between SNPs beyond which the squared correlation coefficient drops below 0.5. Colonies where we found evidence for population structure are show in the column Struct. The multi-dimensional scaling of IBS pairwise distance matrices on which this is based is shown in Figure S1.

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Table 1. Cont.
test this assumption). SNPs at the four loci were spaced so as to allow us to make inferences about both long and short range LD. We assessed the extent of inbreeding and population structure, genetic drift over time, linkage disequilibrium, the proportions of common and rare variants and the extent of genetic differentiation between colonies. Each factor influences the value of a colony for genetic mapping.

Inbreeding and population structure

High rates of inbreeding make colonies less suitable for mapping because they contain fewer segregating QTLs. Table 1 gives four measures of inbreeding: mean minor allele frequency (MAF), heterozygosity (inbred colonies will score low on this measure); the percentage of markers that failed a test of Hardy Weinberg equilibrium (HWE) [11] (colonies that consist of inbred but unrelated individuals, will have high scores) and a coefficient of inbreeding that compares the observed versus expected number of homozygous genotypes [12].

Four colonies are almost inbred (with heterozygosities less than 5%): NTac:NHBS-US, Crl:CD1-IL, Hsd:NIHSBC-IL, BK:W-UK. A further five colonies have heterozygosities between 5% and 10% and so are unlikely to be useful for mapping. Three colonies have inbreeding coefficients greater than 20% (HsdHil: SABRA-IL, Sca:NMRI-SE_10an, HsdOla:MF1-IL) and a further seven have values greater than 10% (Table 1). None of these colonies are suitable for genetic association studies.

Colonies that consist of a mixture of relatives (such as siblings, half siblings, cousins, second degree and third degree relatives) will be difficult to use for mapping because the differing degrees of genetic relatedness introduce population structure. We looked for evidence of this using multi-dimensional scaling of identity by state markers [14] (Fst is a measure of genetic diversity within and among populations [15]).

Low LD is a requirement for high-resolution mapping. We assessed resolution at the four test loci by the LD decay radius, defined as the average physical separation in base pairs between SNPs beyond which the average squared correlation coefficient (r²)

Genetic drift

One potential concern surrounding the use of outbred colonies is that their genetic constitution is not stable and will fluctuate over time, due to unintended directional selection and random genetic drift. Table 1 demonstrates that most colonies are maintained with population sizes of many thousands, which should reduce the effects of shifting allele frequencies. We tested whether this was so by re-sampling six colonies at least one year after our initial analysis and in five cases found good agreement between heterozygosity, relatedness, and inbreeding measured on the two occasions (Table 2). In one case we noted a change in genetic structure. Results obtained from HsdOla:MF1-UK animals used in 2003 were different from those purchased in 2007: heterozygosity fell from 30% to 5% and the inbreeding coefficient rose from 3 to more than 30. Due to infection, the colony had been reformed from a small number of re-derived founders, thereby introducing a severe population bottleneck and explaining the changes in genetic architecture. However, such drastic changes are unusual, are known to the breeders and can be ascertained in advance.

Table 2. Temporal variation.

| Population          | Month | Year | No. | Het. | Pct MAF<5% | Pct fail HWE | Mean inbreeding coef |
|---------------------|-------|------|-----|------|------------|--------------|----------------------|
| Crl:CD1-ICR-US_K64  | Nov   | 2007 | 48  | 0.300| 14.16      | 5.38         | −1.41                |
|                     | Aug   | 2009 | 24  | 0.322| 4.25       | 1.98         | −5.33                |
| Crl:CFW(SW)-US_P08  | June  | 2008 | 206 | 0.216| 24.36      | 0.00         | 4.65                 |
|                     | Oct   | 2009 | 36  | 0.254| 11.33      | 2.83         | 5.29                 |
| HsdIco:OF1-IT       | Nov   | 2007 | 48  | 0.343| 2.27       | 1.36         | 5.22                 |
|                     | Feb   | 2008 | 48  | 0.357| 9.63       | 9.07         | 3.73                 |
| HsdOla:MF1-UK       | Mar   | 2003 | 52  | 0.297| 2.27       | 1.98         | 3.34                 |
|                     | Aug   | 2007 | 192 | 0.051| 1.98       | 31.16        | 31.20                |
| HsdWin:CFW1-NL      | Nov   | 2007 | 48  | 0.205| 7.93       | 4.82         | 3.62                 |
|                     | Aug   | 2008 | 234 | 0.204| 12.18      | 0.00         | 10.19                |
| HsdWin:NMRI-NL      | Aug   | 2007 | 64  | 0.191| 5.95       | 3.12         | 2.11                 |
|                     | Aug   | 2008 | 200 | 0.190| 8.50       | 0.00         | 0.29                 |

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drops below 0.5 (Table 1). Figure 2 shows results for all populations analysed (there were insufficient polymorphic SNPs to calculate LD for NTac:NIHBS-US and ChHli:CD1-IL).

Average figures of LD decay mask variation between regions. For example HsdWin:NMRI-NL has a mean LD decay radius of just over 1, but it will be of little use mapping the MHC region where LD is extensive. However, a region with high LD in one population may have low LD in another. This locus-to-locus variation means that no single population is ideal and that colony-specific genome-wide haplotype and recombination maps are needed.

We explored genome-wide variation in LD in three colonies analysed with the 600K Mouse Diversity Array [16]: Crl:CFW(SW)-US_P08, Crl:NMRI(Han)-FR and Hsd:ICR(CD1)-FR. Mean block length varied between the three colonies: Crl:CFW(SW)-US_P08 79.2 Kb (standard deviation (sd) 70.8), Crl:NMRI(Han)-FR 39.53 Kb (sd 58.7), and Hsd:ICR(CD1)-FR 51.1 Kb (sd 79.5). Block data for each chromosome is given at http://www.well.ox.ac.uk/flint-old/outbreds.shtml. Since there is on average about one gene per 100 Kb, gene-level resolution mapping is possible in these three colonies.

Haplotypes in commercial outbreds are found in laboratory strains

Genome-wide association will be effective in colonies where all, or the majority of haplotypes are tagged by markers on a high-density array. The colonies’ ancestry, as depicted in Figure 1, suggests they contain a relatively limited set of haplotypes, present in inbred strains. We estimated the contribution of each inbred strain to each colony’s genetic architecture by reconstructing the genome of each mouse as a probabilistic mosaic of the founders [17]. We used the Perlegen NIEHS genotypes [18] from 15 inbred strains and analysed all colonies at the four loci (Figure 3) and performed genome-wide analyses in six colonies.

While there is considerable variation between colonies, two general patterns are clear in both locus-specific and genome-wide analyses. First, in all colonies, the fraction of haplotypes accounted for by classical inbred strains ranges between 42% (the NIHS colonies) to 80% (most ICR/CD1). Second, the wild-derived strains (WSB, CAST, MOLF) contribute the least (3–5%). The NIHS stocks contain the highest contribution of the Swiss mouse FVB (25–35%). NMRI are 15–20% FVB and 15% 129, CD1 about 15% FVB and MF1 only 5%. The CFW stocks all contain

![Figure 2. Linkage disequilibrium decay radius (black) and minor allele frequencies (red) in outbred mice. The scale of the vertical axis is megabases for the decay radius and ten times the value of the mean allele frequency (so a value of 2 is 0.2). doi:10.1371/journal.pgen.1001085.g002](image)
about 15% FVB. The genome-wide results are similar, except the overall contribution of 129 is closer to the other classical inbred strains. These results confirm that haplotypes in outbred colonies are predominantly the same as those found in classical laboratory inbred strains and suggest outbred stocks originated from mice genetically similar to inbred strains.

Sequence analysis and novel variants

The haplotype analysis might be subject to SNP ascertainment bias as only variants segregating among inbred strains were genotyped. Furthermore, ancestral haplotype reconstruction always finds representations of the outbred genomes as mosaics of a given set of inbreds; it does not test if the ancestral hypothesis is true in general, nor whether the set of founders is optimal in the sense of explaining the genome structure of outbred mice with the fewest recombinants and inbred strains. However, the ancestral hypothesis would be refuted if many SNPs segregated within the stocks that are not found in inbred strains. Colonies with high rates of these private alleles will be less suitable for genome-wide association studies.

We assessed how many SNPs, missing in laboratory inbred strains, are present in the outbred colonies. We amplified and sequenced 22 fragments of about 1.2 Kb, from eight regions in a 5 Mb region previously sequenced on mouse chromosome 1 [19] and from a further 14 regions within the four QTLs described above. We sequenced 12 animals from three populations.

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Figure 3. Proportion of laboratory inbred strain haplotypes found in commercial outbred stocks. The region above the horizontal black line gives results from an analysis based on 351 markers from four regions in 66 colonies. Below the black line are results from a genome-wide analysis of 6 stocks. The degree of grey scale represents the contribution from each of the Perlegen re-sequenced strains [18] to the outbred colonies. doi:10.1371/journal.pgen.1001085.g003
reads (Crl:CFW(SW)-US_P08 colony, and sequenced sufficient short genome. First, we made a single library from four mice from the account for most of the genetic variation in the colonies tested. A low fraction of novel SNPs suggests that known inbred strains can one in HsdWin:CFW-NL and HsdWin:NMRI-NL (rate 0.8%). The three novel variants (rate 2.5%) in Crl:CFW(SW)-US_K71 and only http://www.sanger.ac.uk/resources/mouse/genomes/). We found C3H/HeJ, C57BL/6N, CBA/J, DBA/2J, LP/J, NOD and NZO derived (129P2, 129S1/SvImJ, 129S5, A/J, AKR/J, BALBc/J, C3H/HeJ, C57BL/6N, CBA/J, DBA/2J, LP/J, NOD and NZO http://www.sanger.ac.uk/resources/mouse/genomes/). We found three novel variants (rate 2.5%) in Crl:CFW(SW)-US_K71 and only one in HsdWin:CFW-NL and HsdWin:NMRI-NL (rate 0.8%). The low fraction of novel SNPs suggests that known inbred strains can account for most of the genetic variation in the colonies tested.

We took two approaches to determine whether these locus-specific results were representative of the rates of SNPs across the genome. First, we made a single library from four mice from the Crl:CFW(SW)-US_P08 colony, and sequenced sufficient short reads (~100 bp) to cover the complete genome at ten-fold coverage. We mapped all reads to the reference genome using MAQ, called SNPs using SAMtools[20,21] and identified a high confidence set of 2,554,879 SNPs. We again compared SNPs with the 13 inbred strains and found that 3.2% of the Crl:CFW(SW)-US_P08 SNPs were novel.

In the second approach, we sequenced libraries of reduced complexity from pooled DNA samples, obtaining high coverage of a small fraction of the genome (~2%). We validated the method by comparing the rate of novel variants found among 36,154 SNPs from a Crl:CFW(SW)-US_P08 reduced-representation library to the rate obtained from our whole-genome sequence described above. 11.7% of the SNPs in the Crl:CFW(SW)-US_P08 sequence were novel. Since the false discovery rate is estimated to be 8% [22], this result implies a novel SNP rate of approximately 4%, consistent with the finding from the whole-genome sequence. We examined four animals from HsdWin:CFW-NL and HsdWin:NMRI-NL colonies and identified 4,885 and 16,724 SNPs respectively. 3.1% of SNPs from HsdWin:CFW-NL and 5.7% of SNPs from HsdWin:NMRI-NL were unique (i.e. not found in the set of SNPs from the inbred strains). These percentages are consistent with there being few, or no novel SNPs in the HsdWin:CFW-NL and HsdWin:NMRI-NL colonies.

Results from genome-wide sequence thus support the conclusions of the locus-specific results: about 95% of the polymorphisms in the colonies are derived from classical inbred strains (those not derived from wild mice). This result is likely true for other colonies, meaning they contain a relatively limited set of haplotypes, consistent with the reconstructions of each mouse as a mosaic of inbred founders described above (Figure 3).

Genetic differentiation between colonies

Our genetic characterization of outbred colonies implies that while the same QTL alleles will segregate in different colonies, their frequencies may vary substantially, so that a QTL segregating in one colony may not be detectable in a second. We assessed the extent of genetic differentiation between colonies and stocks using principal components and Fst distances. We found extensive population differentiation: Fst between populations is 0.454.

No single feature, not stock, colony, producer or country of origin, satisfactorily accounted for genetic differentiation. The top panel of Figure 4 shows the relationships between colonies and the middle panel the relationship between stocks (Figure S2 shows similar results obtained by principal components). We then characterized genetic relationships between colonies regardless of stock identity, using methodologies established in studies of human populations; we considered each colony as originating from K unknown ancestral populations and looked at values of K from 2 to 12 using a maximum likelihood method in the program FRAPPE [23,24]. Figure 4 (bottom panel) shows the results for K = 9 (see http://www.well.ox.ac.uk/flint-old/outbreds.shtml for all values of K). At no value of K were we able to differentiate all stocks.

Stocks differ primarily in the proportions of common ancestry, consistent with their descent from inbred strains. There is considerable variation within a stock, largely explained by variation between colonies, as shown for example by CD1 and NMRI stocks. Taken together the dearth of private alleles, we conclude that quantitative differences in allele frequencies are responsible for population differences.

QTL mapping

From the data in Table 1 we selected three colonies (Crl:CFW(SW)-US_P08, HsdWin:CFW-NL and HsdWin:NMRI-NL) suitable for high-resolution mapping. We mapped four QTLs previously detected in HS mice [6]: serum alkaline phosphatase (ALP) on chromosome 4, the ratio of blood CD4+ to CD6+ T-lymphocytes on chromosome 17, the concentration of high-density lipoproteins (HDL) in serum on chromosome 1, and mean red cell volume (MCV) on chromosome 1. Since HS mice contain alleles derived from eight inbred strains we expect the QTL alleles also to be present in a proportion of the outbred colonies.

We tested first whether QTLs could be detected under the assumption that the QTL alleles descend from inbred progenitors. To do so, we used the ancestral haplotype reconstruction described above and mapped QTLs with the HAPPY software package [17]. The detection of QTLs differed markedly between colonies. There was no evidence for association between any markers on chromosome 1 influencing MCV in any colony (data not shown); single marker association also failed to detect an effect for this phenotype. However, probabilistic ancestral haplotype reconstruction was successful in detecting QTLs for the other three phenotypes. By permutation, we obtained region-specific 5% significance thresholds for HsdWin:NMRI-NL, HsdWin:CFW-NL and Crl:CFW(SW)-US_P08 respectively for ALP of 2.8, 2.4 and 2.6, for HDL of 2.2, 2.9 and 2.6 and for CD4+/CD8+ ratio of 3.3, 2.9 and 2.1, here expressed as a negative logarithm (base 10) of the P-value (logP). Results shown in Figure 5 exceed these thresholds for each phenotype, but not in every colony.

The logP for association with ALP exceeded 2.5 for all colonies in a 400 Kb region between 136.9 Mb and 137.3 Mb on chromosome 4 with considerable variation in the strength of association (logP of 11.5 in HsdWin:NMRI-NL and 2.7 in HsdWin:CFW-NL). One colony showed strong evidence for association with HDL (HsdWin:CFW-NL) with a logP>18; two colonies showed strong association at the chromosome 17 locus with CD4+/CD8+ T-cell ratio (HsdWin:CFW-NL and Crl:CFW(SW)-US_P08). The percentage of variance explained by each QTL is consistent with effect sizes for these phenotypes found in the HS [6]: 15% for CD4+/CD8+ T-cell ratio (in Crl:CFW(SW)-US_P08), 11% for HDL (in HsdWin:CFW-NL) and 18% for ALP (in HsdWin:NMRI-NL).

If the QTL alleles are identical in the three colonies, then a single trait effect for each founder strain, independent of colony, should fit
the data as well as a model in which each colony had independent effects. This allows us to test, for example, whether the QTL alleles influencing ALP in all three colonies are the same. We found that a model for the single trait effect fitted the data as well as one allowing for independent effect. At the peak of association for ALP the P-value of the partial F test was 0.10; for HDL the P-value was 0.27 and for CD4+/CD8+ T-cell ratio, 0.92. Our results indicate that the same QTL alleles are present in the different colonies and confirm that QTL mapping is possible on that assumption.

We then assessed QTL mapping resolution. We wanted to know if our predictions of gene-level mapping (based on estimates of haplotype block length) were upheld in practice. HAPPY mapping results, in Figure 5, indicate a region of over 1 Mb likely to contain each QTL. While this is a smaller region than observed in HS outbred mice (where the mean size of QTL intervals is about 3 Mb) it is larger than suggested by the mean LD decay radius (of about 1 Mb). In fact, the size of the QTL interval is deceptive for two reasons: first, we may not have
modeled the descent from the correct set of progenitors; second, in the absence of fine-scale recombination data, HAPPY mapping assumes a uniform genetic map, without hotspots, so that the localization is relatively imprecise. We resorted therefore to using single marker analysis and considering the LD structure of each region to determine the most likely position of the QTL (Figure 5).

Analysis of the ALP QTL revealed in all colonies a large region of linkage disequilibrium extending from 136.7 to 137.3 Mb, consequently limiting mapping resolution. The region contains an alkaline phosphatase gene (Akp2) at 137.3 Mb, but also an additional 9 genes. Mapping the QTL on chromosome 1 for HDL identified two peaks: rs13476237 at 173631526 and rs3709584 at 173177625 (Figure 5). In the colony showing

Figure 5. Association mapping of three phenotypes in three colonies. The vertical scale is the negative logarithm (base10) of the P-value for the association; the horizontal scale is the position in megabases on the chromosome. On the left of the figure are results for ancestral haplotype reconstruction analysis (HAPPY) for all three colonies; on the right single marker association is shown for one colony for each phenotype: HsdWin:NMRI-NL for alkaline phosphatase; HsdWin:CFW-NL for high density lipoprotein and Crl:CFW(SW)-US_P08 for the CD4+/CD8+ T-cell ratio. LD structure around the associated SNP is shown by a red to white scale for r² = 0 to 1. For high density lipoprotein, each marker is represented by two diamonds. The right hand diamond of each pair is colored to show the r² with rs3476237 (at 173.6 Mb) and the left hand the r² with rs3709584 (at 173.1 Mb). Gene annotations are taken from the UCSC genome browser.

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association (HsdWin:CFW-NL), \( r^2 \) between these markers is low (0.21) and conditioning on the first marker failed to remove the effect attributable to the second \( (F = 15, \text{df} = 2,210, \log P = 6.1) \). These results indicate that two separate effects contribute to the variation in HDL, one co-localizing with \( \text{Apoa2} \), already known to be involved in this phenotype, and the other over a region containing two genes, \( \text{Cd48} \) and \( \text{Slamf1} \), neither previously implicated in the regulation of HDL levels.

On chromosome 17, we found a single peak of association for \( \text{CD4}^+ / \text{CD8}^+ \) ratio at 34.49 Mb (rs33573309). Association with this marker is strongest in the Crl:CFW-US_P08 colony; \( r^2 \) between rs33573309 and rs33699857 at 34550471 is 0.97, but drops to less than 0.3 elsewhere, delimiting a region of 60 Kb containing four genes (Figure 5). Only two of these genes show a strong signal in the joint analysis. \( \text{BCD51142} \) (a.k.a. \( \text{Tesb} \)) is a testis-expressed EST of which little is known; on the other hand, \( \text{H2-Ea} \) encodes the alpha chain of the MHC class II \( \text{E} \) heterodimer, one of the two complexes which govern the selection and survival of \( \text{CD4}^+ \) T cells, and is thus a highly plausible candidate. A number of mouse strains carry a null mutation of \( \text{H2-Ea} \), most often through a 650 bp deletion in the promoter region [25,26], and this deletion is tagged by rs33699857.

We confirmed that the promoter deletion is present in the Crl:CFW(SW)-US_P08 by examining reads from the whole-genome sequence lying between 34,485,333 and 34,483,847 bp [http://www.well.ox.ac.uk/flint-old/outbreds.shtml]. We designed primers to amplify across the deletion and tested for its presence in mice from HsdWin:NMRI-NL and the Crl:CFW(SW)-US_P08 colonies (Figure 6). We performed a complementation analysis to test \( \text{H2-Ea} \), measuring \( \text{CD4}^+/\text{CD8}^+ \) ratios in mice in which the \( \text{H2-Ea} \) null mutation was complemented by introduction of the \( \text{Ea} \) transgene, which drives normal expression of Ea protein with the normal distribution [27]. As illustrated in Figure 6, the presence of the transgene led to an increased representation of \( \text{CD4}^+ \) cells relative to transgene-negative Eanull littermates, confirming the assignment. This increase was present in both the thymus (single-positive mature thymocytes) and spleen, indicating that the variation most likely affects positive selection and lineage commitment of \( \text{CD4}^+ \) T cells.

**Discussion**

Commercially available outbred mice are used primarily by the pharmaceutical industry for toxicology testing, on the assumption that they model outbred human populations, a view supported by limited genetic surveys [28]. In fact, very little is known about their genetic architecture and assumptions about the combined effects of fluctuating allele frequencies (due to genetic drift) and lack of genetic quality control have led some to argue against their use in genetic investigations [29,30]. Our catalogue of the genetic structure of commercially available populations makes a systematic evaluation possible for the first time. Our systematic evaluation of their genetic architecture reveals three important features.

First, variation between colonies is large. \( \text{Fst} \), a measure of variation within and between populations, is 0.454 (in contrast, human populations values are typically less than 0.05 [31]). The source of this variation is not straightforward. Stock names (such as \( \text{NMRI} \) or \( \text{CD1} \)) do not account for it, nor does the supplier, or the country of origin. While some stocks, such as TO and MF1, do indeed have a unique genetic ancestry, many do not. Two likely causes are genetic bottlenecks during colony formation and genetic contamination. Thus, ICR colonies from Harlan and CD-1 colonies from Charles River Laboratories cluster together (Figure 2), having experienced a single bottleneck during their

![Figure 6. A deletion in the promoter of the alpha chain of the MHC class II \( \text{E} \) heterodimer contributes to variation in the ratio of \( \text{CD4}^+ / \text{CD8}^+ \) T-lymphocytes.](https://example.com/figure6.png)
creation (Figure 1). Gene flow appears to have occurred between a number of stocks, as for example between CF1 (HsdWin:CF1) and NMRI (HsdWin:NMRI) colonies of Harlan. Both were bred at the Winkelmann Versuchstierzucht GmbH & Co and could easily have been mixed. A similar story probably explains the close 

and NMRI (HsdWin:NMRI) colonies of Harlan. Both were bred 

creation (Figure 1). Gene flow appears to have occurred between a set of founder strains, shows that the QTLs replicate between 

phenotypic variation in a population will make the trait difficult 

frequency. Large numbers of rare variants contributing to 

colony's suitability for GWAS. Another critical measure is allele 

obtainable in human populations. Applying the same definition of 

outbreds can be modeled as descending from inbred progenitors. 

haplotype reconstruction demonstrates how variants in the 

specific and genome-wide sequencing support this conclusion and 

limited (compared to a wild population). Almost all of the genetic 

resolution genetic mapping (documented in Table 1).

Third, in terms of mapping resolution, no mouse colony is 

comparable to a human population. Using an LD criterion, the 

best mapping resolution in any colony is at least twice that 

obtainable in human populations. Applying the same definition of a 

haplotype LD block as used in human LD studies, we found an 

average block size in three colonies of approximately 60 Kb. By 

contrast, in African populations average block length is 9 Kb, and 

18 Kb in European populations [33].

These observations have important implications for the use of 

commercial outbreds for genetic mapping. First, the predominance of SNPs from classical inbred strains means that arrays 

designed using those SNPs, such as the Affymetrix 600K Mouse 

Diversity Array [16], will capture the majority of genetic variation. 

Second, the extent of LD means that genome-wide coverage can be 

obtained with fewer SNP than in highly outbred and genetically 
heterogeneous populations: using 2 markers to tag each block and 

assuming an average block size of 50 Kb less than 200,000 markers 

will capture the majority of the variation in the genome, so the 

Affymetrix 600K Mouse Diversity Array [16] will be 

adequate. Third, resolution will fall short of gene level in some 

regions. But, since LD structure differs between colonies, high 

resolution mapping of a locus may be possible in one colony, but 

not in another – no single colony is ideal.

However, mapping resolution is not the only useful measure of a colony's suitability for GWAS. Another critical measure is allele frequency. Large numbers of rare variants contributing to phenotypic variation in a population will make the trait difficult to map using standard GWAS designs. In this regard, our data reveal a favorable situation: QTL mapping, assuming a common set of founder strains, shows that the QTLs replicate between stocks in a consistent manner. These findings indicate that quantitative differences in allele frequencies, rather than the existence of private alleles, are responsible for the population differences. Furthermore, the limited sequence diversity means it is possible to impute the sequence of any commercially available mouse from the sequences of inbred strains. Thus, the full catalogue of sequence variation in a stock could be obtained by sequencing the inbred strains presumed to be founders for it, and genotyping the stock at a skeleton of SNPs. Therefore, we should be able to detect the effect of all variants, a situation that has so far eluded studies in completely outbred populations.

Seen in this light, the relatively high degree of genetic differentiation between colonies becomes an advantage. The various genetic architectures available, with variation in QTL frequencies, LD extent and the position of LD blocks, mean that mapping in multiple populations will enable new strategies for gene identification in complex traits. Importantly, we have shown that, at least in the QTLs examined here, the same alleles contribute to variation in different colonies, so that when mapping progress stalls in one stock, another can be used in its stead.

As a proof of principle, we have demonstrated the advantages of mapping in different colonies by detecting the same QTL influencing CD4+/CD8+ ratio and were able to refine this mapping to the gene level, transgene complementation helping to establish as the causal change the deletion in the H2-Ea promoter, a loss-of-function mutation that has long been fixed and segregates widely in the Mus species [26]. A strong genetic influence on the CD4+/CD8+ ratio in mice and human has long been known, predominantly reflecting the efficacy of positive selection [26]. Since MHC class-II molecules such as Ea condition the thymic selection of CD4+ T cells, they are thus highly plausible candidates. The homologous MHC class-II region (HLA-DR) has recently been shown to influence CD4+/CD8+ ratio in human blood [34], providing cross-species validation of our result and an example of how results from mice can inform human genetic studies.

A variety of resources are available for mapping complex traits in mice, each with its own advantages. The choice of which to use depends on the researcher's aims. We advocate commercial outbreds as a resource for finding genes. In some circumstances, as we have shown, it is possible to go from genetic association to a gene in a single step. The sequence variants in commercial outbred colonies are almost solely those present in classical laboratory strains, resulting in three advantages. First, it provides low LD: the colonies do not depend on recombinants accumulated since their foundation. Second, the relatively low genetic diversity increases power to detect a QTL, provided that it segregates, because there will be fewer QTLs overall. Simply put, in a population with ten variants the relative contribution of each is ten times the contribution of each locus in a population with 100 variants. Third, phenotypes known to show heritable variation among the classical laboratory inbred strains will show heritable variation within the outbred colonies.

However, the relatively limited genetic diversity of the outbred colonies means that they do not model a fully outbred population; nor can they be used to assess the effect of all variants present in mouse populations. The colonies contain a relatively small subset of that variation. They are likely to have “blindspots” where little functional variation segregates. The creation of the Collaborative Cross (CC), a large set of recombinant inbred lines derived from genetically diverse progenitors [7], provides access to a more complete catalog of variation [35], and also has the advantage of allowing researchers to interrogate the same genotype multiple times and hence accumulate an increasingly rich understanding of
the relationship between genotype and phenotype. However, it is not yet clear to what extent CC animals will provide high-level mapping resolution, although simulations suggest it will be of the order of 1–2 Mb [36].

Assuming an investigator decides to use an outbred colony, which is the best to choose? For single locus assays, for example attempting to refine a locus identified in a cross between two inbred strains, the choice will depend on whether the appropriate alleles are segregating at the locus, and this can be assessed by haplotype reconstruction from genotype data. The extent of genetic diversity between colonies (ten times that between different human populations) indicates that an appropriate colony will be found. However, genome-wide data will be needed from all colonies to enable a comprehensive assessment. For genome-wide association, which we think is the most likely use of the outbreeds, choice will be guided by the genetic characterization provided here, most simply summarized by low LD, coupled with high mean minor allele frequency. Depending on the phenotype, an additional criterion may be the likelihood that heritable variation is present in a given colony; this could be determined either by family studies carried out with animals from the colony, or by determining whether strains contributing to the colony show phenotypic differences from published data, for example from the phenome project [37]. Our work, demonstrating the utility of the outbreeds, is a starting place for ranking colonies according to their utility for genetic mapping. As costs fall, we anticipate that detailed characterization based on genomic sequence will become available and permit informed choices on the use of the colonies for genetic studies of complex traits in mice.

Materials and Methods

Sample collection and DNA extraction

We contacted ten commercial providers of outbred strain of mice, including Harlan Sprague Dawley (Hsd), Charles River Laboratories (Crl), Taconic Farms (Tac), Centre d’Elevation R. Janvier (Rj), Ace Animal (Aai), B&K Universal (Bk), Hilltop Laboratory Animals (Hla), Research and Consulting Company (Rcc), Scanbur (Sea), Simonsen Laboratories (Sim), and we collected on average 48 tail samples from unrelated mice from each colony (Table S1), representing 90% of all commercially available colonies of outbred mice. 48 unrelated individuals from six colonies were resampled at least one year after the initial collection. We also collected samples from control populations: 109 Collaborative Cross (CC) mice provided by Fuad Iraqi (Tel-Aviv University), 96 DNA samples of wild mice caught in the vicinity of Tucson (Arizona) provided by Michael Nachman, 12 unrelated Heterogeneous Stock (HS) DNA samples from our laboratory and 94 inbred strains purchased from the Jackson Laboratory. DNA was extracted from tail snips using a Nucleopure Kit (Tepnel, UK). DNA quality and quantity was assessed using UV spectrophotometry (Nanodrop) and 0.8% agarose gel electrophoresis.

Genotyping

We designed extension and amplification primers for 351 SNPs using SpectroDESIGNER. Oligonucleotides were synthesized at Metabion (Germany) (Table S1). We used the Sequenom MassARRAY platform for genotyping these 351 SNPs over 4,000 DNA samples and SpectroTYPEr Version 4.1 for data analysis. The resulting genotypes were then uploaded into an Integrated Genotyping System (IGS) [38]. We also obtained genome-wide SNPs genotyping data for six colonies using Affymetrix arrays. Three populations were genotyped using the 600K Affymetrix Mouse Diversity Array [16]. Three more populations were analysed using a precursor to this array, a gift from Mark Daly. DNA was prepared, hybridized and genotypes obtained following the manufacturer’s protocol.

Analyses of genetic relatedness

Data were stored in a relational database designed to manage genotypes and phenotypes [38]. Analyses were run either using software from the authors of each test or were implemented in R [39]. We tested Hardy-Weinberg Equilibrium (HWE) by the exact test [11] for all populations separately. Heterozygosity for each marker was calculated using PLINK [12]. We inferred individual ancestry proportions using a maximum likelihood method [24] in the program FRAPPE (http://www.fhcrc.org/labs/tang/). We used parameters described in [23], running the program for 10,000 iterations, with pre-specified cluster numbers, from K = 2 to 12. We found that independent runs yielded consistent results, with few additional clusters emerging after K = 9. However, it should be noted that given the small set of markers and the inclusion of markers in LD, our estimates of ancestry are likely to be biased. Fit for all pairs of populations was calculated using the FDIST2 program [40,41] (http://www.rubic.rdg.ac.uk/~mab/software.html). An identity-by-state (IBS) matrix for all individuals was calculated using PLINK [12]. Principal component analysis was carried out using this IBS matrix. Genetic relationships were represented as a tree using agglomerative clustering implemented in R [39]. Haplotype blocks were estimated using PLINK [12] which implements the block finding algorithm found in HAPLOVIEW [42].

Big dye sequencing

We used Primer3 to design oligonucleotide primers and carried out PCR reactions with Hotstar Taq obtained from Qiagen. Each 50 μl PCR contained 50 ng of genomic DNA, 1 Unit of HotStar Taq, 5 pmol of forward and reverse primers (synthesized at MWG Biotech, Ebersberg, Germany), 2 mM of each dNTP, 1× PCR buffer and 2.5 mM MgCl₂. We ran the PCR reactions using a Touchdown (TD) approach. The temperature profile consisted of an initial enzyme activation at 95°C for 15 min, followed firstly by 13 cycles of 95°C for 30 sec, 64°C for 30 sec and 72°C for 60 sec, secondly by 29 cycles of 95°C for 30 sec, 57°C for 30 sec and 72°C for 60 sec, and finally by an incubation at 72°C for 7 min. PCR products were purified in a 96-well Millipore purification plate and resuspended in 30 μl of H₂O. Two sequencing reactions were prepared for each DNA sample, one with the forward primer and one with the reverse primer using 50 ng DNA. The sequencing reaction consisted of an initial denaturation stage at 95°C for 1 min, followed by 29 cycles of 95°C for 10 sec, 50°C for 10 sec and 60°C for 4 mins. The PCR reagents were then removed from solution by an ethanol precipitation in the presence of sodium acetate. All sequencing reactions were run out on an ABI3700 sequencer and assembled by using PHRED/PHRAP [43]. Consed was then used for editing and visualisation of the assembly [44].

Short read sequencing

The libraries were prepared from 3–5 μg sample genomic DNA following the Illumina standard genomic library protocol up to the ligation step, where a modified adapter was used. The resulting constructs were digested overnight at 37°C with 20 units high-concentration HindIII restriction enzyme (New England Biolabs) in a volume of 50 μl. The digested libraries were purified on
Qiagen MinElute columns. A complementary biotinylated adapter was ligated to the sticky ends before selecting the fragments of 200 to 500 bp on a 2% agarose gel. The constructs with a HindIII-specific adapter were purified using Streptavidin magnetic beads (Invitrogen) following the manufacturer’s instructions. The beads were finally resuspended in 25 μl 10 mM Tris pH 8.5, of which 12.5 μl were used for the final PCR amplification (15 cycles) using specific amplification primers and Phusion DNA polymerase (Finzymes). The resulting libraries were verified by TOPO cloning and sequencing before running them on an Illumina Genome Analyzer IIx for 38 cycles.

Libraries of reduced complexity for SNP discovery were made from pooled DNA samples. Genomic DNA was subject to complete HindIII restriction enzyme digestion and ligation to linkers. Libraries were then sequenced on an Illumina Genome Analyzer IIx. Since it has been shown that this method has a false SNP discovery rate of about 8% [22], we used three additional filtering criteria to increase confidence in SNP calls. First, following reports that SNPs falling at the ends of reads were unreliable, SNPs within three bases of the end of a read were discarded [22]. Second, SNPs that did not map to within 32 bases of a known HindIII restriction site were also discarded. Third, SNP detection is affected by read depth: where the coverage is less than fivefold, the proportion of novel SNPs rises to over 10%. We only report SNPs where the coverage is greater than tenfold.

Phenotyping

We analysed 200 animals from three colonies: Crl:CFW(SW)-US_P08, HsdWin:CFW-NL, and HsdWin:NMRI-NL. Blood samples were taken from a tail vein and we performed assays for serum alkaline phosphatase (ALP), ratio of CD4+ to CD8+ T-cells, concentration of high-density lipoproteins (HDL) in serum and mean red cell volume using published protocols [45].

Genetic mapping

Where necessary, phenotypes were transformed into Gaussian deviates. Covariates (such as gender, age, experimenter, time) that explain a significant fraction of each phenotype’s variance with ANOVA P-value<0.01 were included in subsequent statistical analyses. We use two mapping methods: a single point analysis of variance of each marker and a multi-point method. The single point method was implemented using linear modeling in R; the multipoint method is implemented in the R package HAPPY [17]. Region-wide significance levels are estimated by permuting the transformed phenotype values 1,000 times.

Supporting Information

Figure S1 Multi-dimensional scaling of identity by state pairwise distances for all colonies, calculated using PLINK. The figure shows a reduced representation of the results, plotting the position on the first dimension (horizontal axis) against position on the second dimension (vertical axis). Found at: doi:10.1371/journal.pgen.1001085.s001 (0.37 MB PDF)

Figure S2 PCA and multi-dimensional scaling of identity by state pairwise distances, calculated using PLINK. The figure shows a reduced representation of the results, plotting the position on the first dimension (horizontal axis) against position on the second dimension (vertical axis). Found at: doi:10.1371/journal.pgen.1001085.s002 (0.37 MB PDF)

Table S1 SNPs used for genotyping the outbred mice Found at: doi:10.1371/journal.pgen.1001085.s003 (0.05 MB XLS)

Table S2 Sequences variants found in outbreds Found at: doi:10.1371/journal.pgen.1001085.s004 (0.22 MB XLS)

Text S1 Origins of commercial outbreds. Found at: doi:10.1371/journal.pgen.1001085.s005 (0.07 MB DOC)

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Author Contributions

Conceived and designed the experiments: BY JN CB RM JF. Performed the experiments: BY JN AB SD JC LF OM AW PK AS DM DJA. Analyzed the data: BY WY XCG MG PK CB DJA. Contributed reagents/materials/analysis tools: BY JN AB SD JC AW. Wrote the paper: BY CB RM JF.

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