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# Gastrointestinal microbiota of wild and inbred individuals of two house mouse subspecies assessed using high-throughput parallel pyrosequencing

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#### **Abstract**

The effects of gastrointestinal tract microbiota (GTM) on host physiology and health have been the subject of considerable interest in recent years. While a variety of captive bred species have been used in experiments, the extent to which GTM of captive and/or inbred individuals resembles natural composition and variation in wild populations is poorly understood. Using 454 pyrosequencing, we performed 16S rDNA GTM barcoding for 30 wild house mice (Mus musculus) and wild-derived inbred strain mice belonging to two subspecies (M. m. musculus and M. m. domesticus). Sequenced individuals were selected according to a 2 × 2 experimental design: wild (14) vs. inbred origin (16) and M. m. musculus (15) vs. M. m. domesticus (15). We compared alpha diversity (i.e. number of operational taxonomic units - OTUs), beta diversity (i.e. interindividual variability) and microbiota composition across the four groups. We found no difference between M. m. musculus and M. m. domesticus subspecies, suggesting low effect of genetic differentiation between these two subspecies on GTM structure. Both inbred and wild populations showed the same level of microbial alpha and beta diversity; however, we found strong differentiation in microbiota composition between wild and inbred populations. Relative abundance of ~ 16% of OTUs differed significantly between wild and inbred individuals. As laboratory mice represent the most abundant model for studying the effects of gut microbiota on host metabolism, immunity and neurology, we suggest that the distinctness of laboratory-kept mouse microbiota, which differs from wild mouse microbiota, needs to be considered in future biomedical research.

Keywords: domestication, hybrid zone, metagenomics, microbiome, Mus musculus, symbiosis Received 28 July 2013; revision received 2 September 2014; accepted 3 September 2014

#### Introduction

Vertebrates interact with bacterial communities that are both specialized and highly diversified taxonomically and spatially (Costello *et al.* 2009). Bacterial cell and active gene counts for animal-associated microbiota exceed the number of host somatic cells and genes by at least an order of magnitude (Qin *et al.* 2010). It is not

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surprising, therefore, that bacterial communities have a marked influence on the physiology and health of the host (Sekirov *et al.* 2010; Wallace *et al.* 2011).

Gastrointestinal tract microbiota (GTM) have been the most heavily studied vertebrate-associated bacterial community (De Filippo *et al.* 2010; Qin *et al.* 2010; Sekirov *et al.* 2010; Foster & Neufeld 2013). Functionally, GTM are often viewed as a separate metabolic organ with many causal pleiotropic effects on the host (O'Hara & Shanahan 2006). Studies using gnotobionts and germ-free animal models (i.e. animals lacking GTM) have shown

that the presence of GTM is essential for normal development of the digestive tract (Wostmann & Bruckner-Kardoss 1959; Gordon & Bruckner-Kardoss 1961) and the immune (Macpherson & Harris 2004) and central nervous systems (Heijtz *et al.* 2011; Cryan & Dinan 2012). GTM also play an important role in host energy balance (Hooper *et al.* 2001; Bäckhed *et al.* 2005, 2007; Sonnenburg *et al.* 2005; Jumpertz *et al.* 2011). Consequently, GTM dysbiosis can lead to pathologies such as obesity (Turnbaugh *et al.* 2006), autoimmune diseases (Mathis & Benoist 2011), pathogen invasions (Koch & Schmid-Hempel 2011; Chervonsky 2012) and neurological disorders (Cryan & Dinan 2012).

At higher taxonomic levels, GTM are typically dominated by two bacterial phyla, the Bacteroidetes and Firmicutes (Ley *et al.* 2008; Ochman *et al.* 2010; Muegge *et al.* 2011). At the phylotype (i.e. operational taxonomic unit; OTU) level, interindividual differences are much more pronounced, with the majority of taxa being unique to a given individual (Turnbaugh *et al.* 2009; Qin *et al.* 2010). It is possible, however, to identify a group of bacterial taxa that are shared between individuals and that enable the key functions of the GTM (Turnbaugh & Gordon 2009; Jalanka-Tuovinen *et al.* 2011).

Factors contributing to high levels of interindividual difference in GTM composition are still the subject of intensive research, as well as severe controversy. Maternal transfer during the early stages of ontogeny has a significant long-lasting effect on GTM composition (Lucas & Heeb 2005; Turnbaugh *et al.* 2009; Friswell *et al.* 2010). In addition, GTM composition is influenced by environmental conditions during early life (Sanchez *et al.* 2011) and diet composition (Faith *et al.* 2011; Wang *et al.* 2014). Finally, it has also been suggested that genetic background has an effect on GTM structure (Rawls *et al.* 2006; Benson *et al.* 2010).

At the population level, GTM profiles tend to cluster together based on the geographic origin of the host (De Filippo *et al.* 2010; Nam *et al.* 2011; Banks *et al.* 2009; see also Wang *et al.* 2014). Geographic distance has also been found to be the most significant factor explaining patterns of diversity in the GTM of house mice, with host population structure and genetic distance having a comparatively weaker influence (Linnenbrink *et al.* 2013).

The effect of host phylogeny on GTM composition is not yet completely resolved. During mammalian evolution, GTM composition appears to have been unstable and shaped through independent acquisitions of bacterial taxa in animals with convergent dietary strategies (Ley *et al.* 2008; Muegge *et al.* 2011).

The wild house mouse (*Mus musculus*) is a polytypic species often used as a model species in population and evolutionary biology (Macholán *et al.* 2012). In

particular, the European hybrid zone between the M. m. musculus (hereafter Mmm) and M. m. domesticus (Mmd) subspecies is an important model for studying the evolution of speciation (Baird & Macholán 2012; Dureje et al. 2012). Limited hybridization between the two subspecies has been put down to various reproductive barriers (Dufková et al. 2011; Duvaux et al. 2011; Macholán et al. 2011; Vošlajerová-Bímová et al. 2011; Albrechtova et al. 2012; Janousek et al. 2012; Turner et al. 2012). While the genetic component of decreased hybrid fitness has frequently been studied (reviewed in Forejt et al. 2012; Dzur-Gejdosova et al. 2012; Flachs et al. 2012), studies of the epigenetic component of reproductive isolation are rather scarce (see for example Schütt et al. 2003), and possible differences in gut microbiota composition between mouse subspecies have not been considered at all. Nevertheless, previous studies on nonmammalian species suggest that differences in GTM composition may contribute to assortative mating and promote reproductive barriers (Sharon et al. 2010).

Laboratory strains of the house mouse are the most widely used animal model for biomedicine and human genetics (Fox et al. 2007), including research focused on metabolic and physiological effects of GTM on the host (reviewed in Sekirov et al. 2010; Staubach et al. 2012). However, the diversity and genetic composition of laboratory strains are not truly representative of wild house mice (Yang et al. 2011). This is mainly due to a founder effect (i.e. most laboratory strains are of Mmd origin), inbreeding and, sometimes, the hybrid origin of strains. It is probable that similar processes may also have influenced GTM composition in laboratory mice, and hence 'laboratory GTM' will not represent 'wild GTM'. Differences between laboratory and wild GTM may become further pronounced through the exposure of these two groups to completely different environmental conditions (e.g. temporal and spatial energy intake or presence of pathogens and stress; Sudakaran et al. 2012; Nelson et al. 2013). Interindividual variability in GTM composition in wild populations vs. inbred strains is also of interest as this can strongly influence effect size estimates and the significance of experimental results. While knowledge of GTM composition in classical laboratory mouse strains is growing rapidly (Benson et al. 2010; Friswell et al. 2010; Krych et al. 2013), it is still lacking for wild-derived mouse strains with welldefined breeding histories and genetic origin. There has been just one study of GTM composition in wild house mice, undertaken by Linnenbrink et al. (2013). A direct comparison of GTM between wild populations and captive bred strains, therefore, has yet to be performed.

In this study, we perform 16S rRNA barcoding, to assess gut microbiota composition in wild and wild-derived inbred individuals of Mmm and Mmd origin.

We evaluate interindividual variation and differences in GTM diversity and composition between wild populations and inbred mouse strains, as well as between the two subspecies. Our results allow us to address questions regarding the role of genetic background vs. environmental conditions on diversity and interindividual variation in GTM. We also used 16S rRNA sequences to predict GTM functional potential (i.e. genes encoded by GTM) based on the recently developed algorithm of Langille *et al.* (2013). This enabled us to gain preliminary insights into metagenomic variation between Mmm and Mmd and between wild vs. wild-derived inbred individuals.

#### Materials and methods

# Mouse samples

Each of the 14 wild mice used in this study was captured at a different locality along the Czech-Bavarian border region (length of east–west transection = 142 km; Fig. 1). Subadult mice and mice in bad condition were not included in the study. In addition, a male and female from each of eight wild-derived inbred strains kept at the conventional breeding facility of the Studenec Research Facility (Institute of Vertebrate Biology, Academy of Sciences, Czech Republic) were divided into pairs based on their source localities. Details of Mmm strains BUSNA, BULS (source locality Buskovice) and STUF, STUS (Studenec) and Mmd strains STRA, STRB (Straas) can be found in Piálek et al. (2008); the two remaining Mmd strains, SCHEFE and SCHUNT, originated from Schweben in Central Germany [50°26'N, 9°35'E]. Thus, the 16 inbred mice had their origin in eight inbred strains from no more than 4 wild localities, giving a high probability of decreased genetic diversity due to a founder effect. See Table S1 (Supporting information) for more details on individual samples.

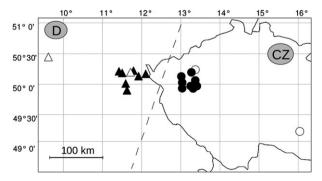


Fig. 1 Sampling localities for wild (full symbols) and inbred (open symbols) populations of *Mus musculus musculus* (circles) and *Mus musculus domesticus* (triangles). The dashed line represents the house mouse hybrid zone.

The wild-derived strains were kept in 16 cm×28 cm×15 cm Perspex cages with sawdust bedding under the following constant conditions: a 14:10 light:dark cycle, 22 °C temperature, 40% humidity. Standard mouse pellets (ST1, VELAZ) and tap water were available *ad libitum*. The breeding facility is licensed for keeping small mammals according to Czech law (3OZ31162/2011-17214).

# Laboratory procedures

Wild mice were live-trapped and caged singly until dissection in a field laboratory. Sterile dissection was performed no more than 24 h after the mouse was trapped. Colons, including faeces, were stored in 70% EtOH at room temperature. DNA was isolated from the faeces at the Institute of Parasitology, Academy of Sciences, Czech Republic, as described in Sak *et al.* (2011). Inbred individuals were euthanized and dissected at the breeding facility in Studenec. Faeces were placed into extraction buffer and DNA extracted using the same commercial kit as for wild mice (i.e. QIAamp® DNA Stool Mini Kit, QIAGEN). All DNA samples were diluted to a concentration of 5 ng/μL (Qubit fluorometer, Invitrogen).

To perform bacterial barcoding, we used MPRK341F (CCTAYGGGRBGCASCAG) and MPRK806R (GGACT ACNNGGGTATCTAAT) primers to amplify the 466 bp fragment consisting of the V3 and V4 regions of Escherichia coli 16S rDNA (Yu et al. 2005). Both the fusion primers needed for 454 sequencing and the amplification protocol were kindly provided by Lars Hansen (University of Copenhagen). The reverse fusion primer consisted of the Titanium adaptor A sequence (Lib A) and the MPRK806R primer sequence. Forward fusion primers differed between the individuals sequenced within a single sequencing run and consisted of adaptor B sequence (Lib A), a unique tag sequence from the Roche MID library and the MPRK341F primer sequence. 16S rDNA was amplified in two independent PCRs. A first PCR was performed with the MPRK341F and MPRK806R primers twice for each sample and the two PCRs pooled. The PCR mix (20 µL) consisted of 0.02 U/µL Phusion Hot Start DNA Polymerase II (Finnzymes), 1× Phusion HF buffer with MgCl2, 0.2 mm dNTP, 0.5  $\mu M$  of each primer and 0.5  $\mu L$  of DNA. The PCR conditions consisted of an initial denaturation at 98 °C for 30 s, followed by 32 cycles of 98 °C (5 s), 56 °C (20 s) and 72 °C (20 s), and a final extension at 72 °C (5 min). PCR products were incubated at 70 °C for three minutes and moved to ice. The samples were run on 1% agarose gel, PCR bands were excised from the gel, purified using the QIAquick gel extraction kit (QIAGEN) and eluted in 30 µL of elution buffer. A second PCR was performed to add sequencing adaptors and sample specific tags to the amplicons. This PCR was performed in 40  $\mu L$  and was the same as the previous one, except that fusion PCR primers were used, 1  $\mu L$  of the first PCR was used as a template and the number of PCR cycles was cut down to 15. PCR products were warmed (70 °C/3 min) and incubated on ice, separated onto agarose gel, excised from the gel and purified as described above, and their concentration was measured using PicoGreen (Invitrogen) and normalized.

Amplicons were then pooled (separate pools for wild and for inbred samples) and pyrosequenced using GS Junior System (Roche) in two independent sequencing runs according to the manufacturer's instructions. To achieve unidirectional sequencing starting from the tagged forward primers, only B beads were used in the emulsion PCR.

## Data analysis

Ampliconnoise pipeline (Quince *et al.* 2011) was used for denoising and quality and chimera filtering of the original sff files. The resulting high-quality sequences were further clustered into OTUs based on 97% sequence similarity, using the CLUSTOM algorithm (Hwang *et al.* 2013).

CLUSTOM outputs were converted to QIIME input format using in-house R scripts and further processed using QIIME pipeline (Caporaso *et al.* 2010a), whereby OTU taxonomy was assigned using the RDP classifier (Wang *et al.* 2007; confidence to record assignment >0.80). Representative sequences were aligned using PyNast and Greengenes Core Set Alignment (Caporaso *et al.* 2010b). A minimum evolution phylogenetic tree was constructed using FASTTREE (Price *et al.* 2009).

To assess variation in alpha diversity between individual subspecies (Mmm vs. Mmd) and between individuals from wild populations vs. inbred strains, we used rarefaction analysis (10 rarefied subsamples of sequences for each individual and sequence counts per individual sample ranging between 100 and 3500). Three diversity measures were computed on the subsampled data: (i) the number of OTUs in a given rarefied subsample, (ii) the Chao1 index (Chao & Shen 2003), and (iii) a phylogenetic diversity index computed as total branch length.

We further computed weighted and unweighted UNI-FRAC distances (Lozupone & Knight 2005) between individuals and visualized the overall pattern of interindividual variation using Nonmetric Multidimensional Scaling (NMDS). To avoid potential bias associated with unequal sequencing depth in the samples, UNI-FRAC analysis is based on a random subsample

corresponding to 1200 reads (~minimum achieved sequencing depth) per individual. Next, we assessed to what extent (i) subspecies identity (Mmm vs. Mmd), (ii) mouse origin (wild vs. inbred), and (iii) the interaction between these variables contribute to variation in microbiome composition. To do this, we used multivariate ANOVA based on dissimilarities (ADONIS) implemented in the R (v. 2.14.1, R Development Core Team 2011) VEGAN package (Oksanen et al. 2013). The BETADISPER function (analogous to Levene's test) was used to test whether the interindividual variation in UNIFRAC distances was the same for samples originating from different subspecies and for inbred vs. wild individuals. ADONIS was further used to test for the effect of source population on GTM composition in inbred strains. Using the Mantel test, we assessed the correlation between UNIFRAC distances and geographic distance in wild individuals.

Differences in taxonomic composition between subspecies and inbred vs. wild individuals at the phylum level were further evaluated using the Mann–Whitney *U*-test and the Šidak multiple test correction method. The same comparison based on Welsh's t test provided identical results (not shown).

Further, we used permutation ANOVAS to analyse variation in taxonomic composition at the level of individual OTUs. We chose an arbitrary OTU threshold of those with >50 high-quality sequences (n = 81) as these OTUs represented the vast majority (95.7%) of all high-quality reads. First, we converted read counts for individual OTUs within sampled individuals into proportions. We further generated 1000 randomized community matrices by resampling (without replacement) OTU proportions across samples. Using custom R scripts, separate factorial ANOVA models were fitted for individual OTUs and for both original and resampled community matrixes, including sample origin (wild vs. inbred), species identity (Mmm vs. Mmd) and their interactions as explanatory variables. Significance of these variables for a given OTU was assessed based on a comparison of t values computed for original data with the distribution of simulated t values (Lepš & Šmilauer 2003). The Q value method for multiple tests correction (QVALUE software; Storey & Tibshirani 2003) was used to minimize the risk of false discovery rates.

PICRUST pipeline (Langille *et al.* 2013) was used to predict gene content of individual GTM samples. In short, this software uses 16S RNA data on taxonomic composition of a microbiome, together with information on bacterial gene content that have been whole-genome-sequenced and an ancestral-state reconstruction algorithm to predict functional content (in terms of encoded genes) of given community. First, we mapped our high-quality sequences against reference OTUs from the Green Gene reference database (version:

gg\_13\_5\_otus; DeSantis *et al.* 2006; 97% similarity threshold) using the closed reference algorithm implemented in QIIME. The resulting OTU table was used as the PICRUST input. PICRUST was run using the default setup recommended by the authors. The resulting database included information on predicted abundance of individual Kyoto Encyclopedia of Genes and Genomes (KEGG) pathways (Kanehisa & Goto 2000) in individual samples. This data set was further analysed using permutation-based ANOVAS, as described above.

#### Results

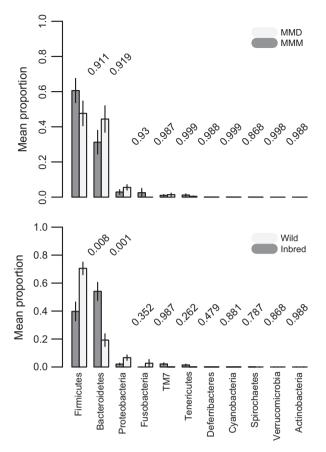
#### General description of house mouse GTM

We identified 107999 sequences that passed quality filtering and did not correspond to chimeric OTUs, with the number of high-quality sequences per individual ranging between 1230 and 6811 (mean = 3599). These sequences were clustered in 1374 OTUs based on 97% sequence similarity. The RDP classifier successfully classified 865 OTUs (which included 10 336/>95%/ high-quality sequences) to family level. RDP failed to classify just 157 OTUs to phylum level. These taxonomically unclassified OTUs, however, represented just a small minority (~0.5%) of the high-quality sequences.

The house mouse microbiome was dominated by bacteria of the phyla *Firmicutes* and *Bacteroidetes*. Other bacterial phyla identified included *Proteobacteria*, *Fusobacteria*, *TM7*, *Tenericuides*, *Deferribacteres* and 10 others. Compared to *Firmicutes* and *Bacteroidetes*, however, their relative abundances were low (see Fig. 2; details on taxonomic composition is available in Appendix S1 and Table S2, Supporting information).

# GTM diversity

The number of OTUs per individual, as predicted by the Chao1 index, ranged between 650 and 354, with the exception of one inbred outlier with a considerably lower alpha diversity level (Table S1, Supporting information). Rarefaction analysis suggested that the sequencing depth was insufficient to capture all GTM diversity (Fig. 3); nevertheless, rarefaction curves indicated no pronounced difference in alpha diversity between Mmm and Mmd (mean Chao1  $\pm$  SE =  $390 \pm 40$  and  $320 \pm 27$ , respectively) or between samples from inbred strains and wild population (mean Chao1  $\pm$  SE = 359  $\pm$  32 and 350  $\pm$  38, respectively; Fig. 3; Table S1, Supporting information). We tested the effect of these variables, and their interaction, on observed alpha diversity using linear mixed effect models (fitted using R package lme4; Bates et al. 2012). Alpha diversity indices computed for rarefied GTM



**Fig. 2** Differences in taxonomic composition at the phylum level, expressed as mean proportion of sequences in *Mus musculus musculus* (MMM) vs. *Mus musculus domesticus* (MMD) individuals (upper graph) and wild vs. inbred individuals (lower graph). Error lines indicate standard errors for individual groups. *P* values above bars correspond to Welsh's *t*-tests and the Sidak multiple test correction method.

subsamples (*n* = 10 for each individual and each sequence count category), including 1000 (~minimum sequencing depth) and 3000 (~median sequencing depth) randomly selected sequences, were used as response variables. Identity of individuals, corresponding to the replicated rarefied subsamples, was included as a random intercept. None of these explanatory variables had a significant effect on observed alpha diversity (Table 1), although rarefaction plots for number of observed OTUs and phylogenetic diversity suggest slightly decreased alpha diversity for inbred individuals (see Fig. 3 and Appendix S2, Supporting information).

# Variation in community structure

The NMDS plot based on both weighted and unweighted UNIFRAC distances revealed differences in faecal microbiome composition between individuals originating from inbred individuals and wild populations

**Table 1** Effect of subspecies identity (Mmm vs. Mmd), origin (inbred vs. wild) and interaction between these variables on three alpha diversity measures (i.e. number of OTUs, Chao1 index and Phylogenetic diversity index) for rarefied subsamples of the original data set. n = 1000 and 3000 randomly selected sequences per individual. Data were analysed using linear mixed effect models. Significance of explanatory variables was assessed using the likelihood ratio test (e.g. Crawley 2007)

Diversity index	Explanatory variable	d.f.	n = 1000		n = 3000	
			$\chi^2$	P	$\chi^2$	P
Number of OTUs	Subspecies	1	1.131	0.288	0.015	0.903
	Origin	1	1.116	0.291	0.027	0.869
	Subspecies × Origin	1	0.028	0.868	0.010	0.993
Chao1	Subspecies	1	0.066	0.797	3.114	0.078
	Origin	1	0.219	0.640	2.803	0.094
	Subspecies × Origin	1	0.112	0.738	2.945	0.086
Phylogenetic diversity	Subspecies	1	1.540	0.215	0.184	0.668
	Origin	1	0.195	0.195	0.682	0.409
	Subspecies × Origin	1	0.010	0.922	0.136	0.712

(Fig. 4). We detected one outlying point in the NMDS plot that exhibited decreased alpha diversity (corresponding to sample 'I' in Table S1, Supporting information). Subsequently, ADONIS and BETADISPER models were evaluated for distance matrices that included (presented in Appendix S2, Supporting information) and excluded (presented in this section) this sample. If not explicitly stated, both options provided consistent results.

In accordance with the NMDS plots, ADONIS confirmed significant differentiation between inbred and wild individuals. Conversely, ADONIS failed to detect significant differences between Mmm and Mmd. The effect of interaction between species identity and origin was significant in the case of weighted UNIFRAC, but not in the case of unweighted UNIFRAC (Table 2).

Separate ADONIS analyses for wild and inbred samples do not suggest GTM differentiation between Mmm and Mmd (Wild samples:  $F_{(1,12)} = 0.903$ ,  $R^2 = 0.070$ , P = 0.424 and  $F_{(1.13)} = 0.768$ ,  $R^2 = 0.060$ , P = 0.654, Inbred samples:  $F_{(1,13)} = 1.042$ ,  $R^2 = 0.070$ , P = 0.366 and  $F_{(1.13)} = 1.186$ ,  $R^2 = 0.080$ , P = 0.194, for weighted and unweighted UNIFRAC, respectively). On the other hand, we observed significant differentiation between wild and inbred Mmm individuals using weighted UNIFRAC ( $F_{(1.12)} = 12.868$ ,  $R^2 = 0.520$ , P = 0.003), while the results were nonsignificant in the case of Mmd  $(F_{(1.14)} = 1.448, R^2 = 0.100, P = 0.223)$ . Unweighted UNI-FRAC revealed significant differentiation between inbred and wild samples for both Mmd and Mmm  $(F_{(1,13)} = 4.259, R^2 = 0.250, P = 0.002 \text{ and } F_{(1,13)} = 2.650,$  $R^2 = 0.180$ , P = 0.004, respectively).

Using ADONIS, we found no significant effect on GTM composition of source population used for establishment of individual strains (weighted UNIFRAC:  $F_{(3.12)} = 1.259$ ,  $R^2 = 0.230$ , P = 0.144 and unweighted

UNIFRAC:  $F_{(3,12)} = 1.402$ ,  $R^2 = 0.250$ , P = 0.234) or sex (weighted UNIFRAC:  $F_{(1,14)} = 0.692$ ,  $R^2 = 0.050$ , P = 0.882 and unweighted UNIFRAC:  $F_{(1,14)} = 0.842$ ,  $R^2 = 0.060$ , P = 0.472).

BETADISPER revealed no difference in interindividual variation in GTM composition between inbred and wild individuals (weighted UNIFRAC:  $F_{(1,27)} = 0.300$ , P = 0.598, unweighted UNIFRAC:  $F_{(1,27)}$ ,  $R^2 = 0.568$ , P = 0.457) or between Mmm and Mmd (weighted UNIFRAC:  $F_{(1,27)}$ ,  $R^2 = 0.128$ , P = 0.719, unweighted UNIFRAC:  $F_{(1,27)}$ ,  $R^2 = 0.335$ , P = 0.568).

The correlation between geographic distance between wild specimens and dissimilarity indices for their GTM was not significant (Mantel test; weighted UNIFRAC: r = -0.0134, P = 0.506 and unweighted UNIFRAC: r = -0.0684, P = 0.759).

#### Variation in the abundance of taxonomic signatures

Significant differences were detected in the proportion of reads assigned to the *Bacteroidetes* and *Firmicutes* phyla between wild and inbred individuals using the multiple Mann–Whitney *U*-test. The proportions of other phyla did not differ significantly between wild and inbred individuals. In addition, we found no significant variation in taxonomic composition between Mmm and Mmd at the phylum level (Fig. 2).

Permutation-based ANOVAS did not support variation in the relative abundance of individual OTUs between Mmm and Mmd, with OTUs represented differently in the two subspecies. Conversely, we observed pronounced differences between wild and inbred individuals, with 16% of OTUs varying significantly. Specifically, OTUs corresponding to the *Lachnospiraceae*, *Lactobaciallaceae*, *Ruminococaceae* and *Porphyromonadaceae* families were more abundant in wild individuals and

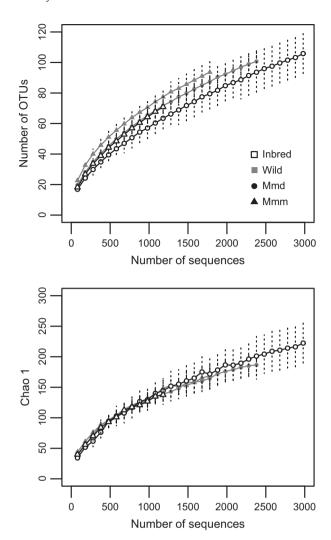
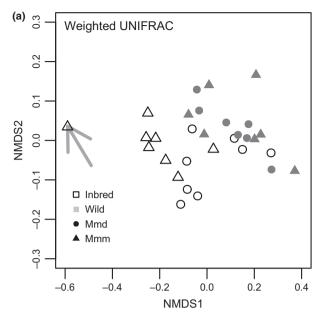
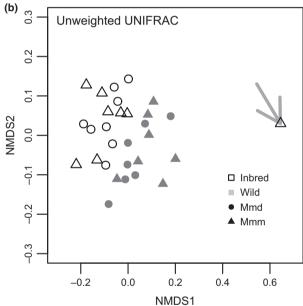


Fig. 3 Mean number of OTUs and Chao1 index (± SE) per number of reads detected in *Mus musculus musculus* (circles) and *Mus musculus domesticus* (triangles), and wild (full symbols) and inbred (open symbols) individuals. Rarefaction curves presented only for the range that does not exceed minimum number of high-quality sequences per individual detected within a given category.

*Rikenellaceae* and *Bacteroidaceae* in inbred individuals. Finally, there was low support for an interactive effect between sample origin and subspecies identity, with a significant interactive effect being detected in a single OTU (Table S2, Supporting information).

Permutation ANOVAS were performed separately for individual subspecies. Following q-value adjustment, we detected four and one OTUs differing significantly between wild vs. inbred individuals in Mmm and Mmd samples, respectively. The standardized effect sizes for individual OTUs were positively correlated between Mmd and Mmm ( $F_{(2,77)} = 27.180$ ,  $R^2 = 0.250$ , P < 0.001).





**Fig. 4** Nonmetric multidimensional scaling plots for (a) weighted and (b) unweighted UNIFRAC distances for wild (full symbols) and inbred (open symbols), and *Mus musculus musculus* (circles) and *Mus musculus domesticus* (triangles) individuals. Outlying sample is indicated by arrow.

#### Metagenomic predictions

We were unable to map 9104 sequences (8.5% of high-quality reads) to GreenGenes reference database, when constructing input OTU table for PICRUST. The proportion of unmappable sequences did not differ dramatically between wild (n = 3499; 7.4%) and inbred samples (n = 5605; 9.3%). ADONIS analysis based on Bray–Curtis distances for individual metagenomes revealed significant

**Table 2** Results for multivariate ANOVA based on dissimilarities (ADONIS). Weighted and unweighted UNIFRAC distance matrices were used as independent variables. Variables with a significant effect are in italics. The outlier from Fig. 2 was not considered in this analysis; its exclusion, however, did not change the results (for more details, see Appendix S2, Supporting information)

Distance measure	Explanatory variable	d.f.	MSS	F	$R^2$	Р
Weighted UNIFRAC	Origin	1	0.162	7.923	0.211	0.003
	Subspecies	1	0.032	1.573	0.042	0.180
	Subspecies × Origin	1	0.063	3.090	0.082	0.043
	Residuals	25	0.020		0.665	
Unweighted UNIFRAC	Origin	1	0.017	6.781	0.199	< 0.001
	Subspecies	1	0.003	1.162	0.034	0.277
	Subspecies × Origin	1	0.003	1.074	0.032	0.334
	Residuals	25	0.002		0.735	

differences in the functional potential of the wild vs. inbred group ( $F_{(1.26)} = 7.349$ ,  $R^2 = 0.200$ , P < 0.005). Differences between Mmd and Mmm, as well as the interactive effect of subspecies identity vs. sample origin, were not significant  $(F_{(1,26)} = 1.842, R^2 = 0.050, P = 0.153)$  and  $F_{(1.26)} = 2.061$ ,  $R^2 = 0.060$ , P = 0.130, respectively). We repeated this analysis using z-score transformed relative abundances of individual KEGG categories to compute a Euclidean distance matrix as the ADONIS input. This normalization caused all KEGG categories to have the same weight during analysis, irrespective of their relative abundances in predicted metagenomes. Despite this, the ADONIS results remained unchanged, revealing highly significant differences between wild and inbred samples  $(F_{(1,26)} = 6.116, R^2 = 0.170, P < 0.001)$  and no effect of subspecies identity ( $F_{(1,26)} = 1.946$ ,  $R^2 = 0.050$ , P = 0.077) or interaction of these two variables ( $F_{(1,26)} = 1.620$ ,  $R^2 = 0.040$ , P = 0.130). Subsequently, PICRUST analysis, followed by permutation-based tests, predicted >35% KEGG categories to be significantly over- or under-represented in wild compared to inbred individuals, whereas the abundances of only ~2% KEGG categories varied between Mmm and Mmd and ~6.5% KEGG categories changed interactively according to subspecies vs. origin identity (Table S3, Supporting information).

# Discussion

Taxonomic assignment of 16S rRNA sequences indicated that the house mouse GTM consists of six bacterial phyla and is dominated by *Bacteroidetes* and *Firmicutes*. This is in agreement with previous studies on GTM composition in wild mice of Mmd origin (Linnenbrink *et al.* 2013), laboratory mouse strains (Staubach *et al.* 2012; Ubeda *et al.* 2012) and most mammalian taxa studied so far (Ley *et al.* 2008; Muegge *et al.* 2011). GTM diversity ranged between 131 and 633 OTUs per individual, as predicted using the Chao1 index.

Our GTM data set benefits from the fact that samples from wild Mmm and Mmd populations were obtained at a small spatial scale, resulting in a lack of correlation between geographical distance and GTM similarity among individuals (as assessed using Mantel's test). Hence, the comparison of GTM structure between the two subspecies is unlikely to be affected by confounding factors, such as clinal variation in environmental conditions (e.g. Banks *et al.* 2009; Linnenbrink *et al.* 2013). Consequently, our data allow direct testing of the hypothesis that genetic differences alone underline potential GTM diversification between Mmm and Mmd.

We found that both Mmm and Mmd subspecies harbour the same level of GTM alpha diversity. In addition, interindividual variation (i.e. beta diversity) in GTM composition between both subspecies was comparable. This implies a low effect of host genetic diversity on GTM structure, given the higher genetic diversity and higher expected interindividual variation in Mmm genetic background compared to Mmd (Geraldes *et al.* 2008; Phifer-Rixey *et al.* 2012).

There is growing evidence that the host genome interacts with the GTM community and shapes its taxonomic composition (Rawls et al. 2006). Nevertheless, we found no pronounced difference in GTM taxonomic composition between Mmd and Mmm mice. This indicates a limited effect of both separate evolutionary history and differences in life history traits and ecology between Mmm and Mmd (Macholán et al. 2012) on the functional divergence of genes underlining GTM content. It is worth noting that weighted UNIFRAC analysis indicated a significant interactive effect of subspecies identity and captive vs. wild origin on GTM composition. Separate analyses indicated a higher difference in wild vs. inbred microbiota for Mmm than for Mmd suggesting a pronounced response to captive breeding in Mmm. Nevertheless, the relative abundance of individual OTUs varied between wild and captive individuals in a similar way in both house mouse subspecies,

suggesting that the GTM of Mmm and Mmd tend to exhibit a congruent response to captive conditions.

Previous experimental studies have shown that differences in bacteria inhabiting *Drosophila* result in a strong assortative mate choice, which may contribute to the establishment of precopulatory reproductive barriers (Sharon *et al.* 2010). Our data, however, suggest that assortative mating based on GTM composition is unlikely to contribute to the reproductive isolation of Mmm and Mmd as GTM composition is almost identical in both Mmm and Mmd mice, at least in wild populations.

It is to be expected that both alpha (sensu OTU richness) and beta (sensu interindividual variation) GTM diversity will be higher in wild populations compared to wild-derived inbred strains. First, inbred mice in a breeding facility are kept solely on a standardized diet, while the diet of wild animals is likely to exhibit much higher temporal and spatial variation. According to niche differentiation and resource partitioning theory (Gause 1932; MacArthur 1958), which assumes a link between environmental heterogeneity (variability of diet in this particular context) and community richness (GTM composition in our case), diet uniformity should lead to low GTM diversity in inbred strains. Secondly, a decrease in genetic diversity in host populations due to a founder effect and inbreeding may be manifested in decreased GTM beta diversity in inbred strains if the host genotype has a direct effect on GTM composition. Thirdly, decreased GTM beta diversity in inbred strains is expected due to the sampling of GTM from fewer source localities for inbred strains compared to wild populations. Finally, limited social contact between captive individuals compared to free interaction in wild individuals is likely to result in decreased GTM beta diversity in inbred strains due to limited horizontal transfer of bacteria (Koch & Schmid-Hempel 2011). Our results, however, oppose these expectations as alpha and beta GTM diversities in both inbred strains and wild population were the same. Interestingly, Nelson et al. (2013) recently found GTM alpha diversity of southern elephant seals (Mirounga leonina) to be even higher in captive breeds compared to wild populations. What is more, the number of sequences that were successfully mapped against the database of annotated reference 16S rRNA sequences was high and comparable for both wild and inbred individuals (92.6% and 90.7% in the case of wild and inbred samples), suggesting that the proportion of unknown or poorly described microbial diversity is unexpectedly low in wild house mouse populations.

Despite comparable levels of GTM diversity between wild and wild-derived inbred populations, GTM taxonomic composition varied markedly between these two categories. To our knowledge, there have been only few studies focusing on differences in GTM composition in wild populations vs. captive individuals (Sudakaran et al. 2012; Nelson et al. 2013; Kohl & Dearing 2014). Most of them observed pronounced differentiation between the two categories. Similarly, a previous study on laboratory mouse GTM (Friswell et al. 2010) showed that individuals tended to be clustered together based on origin (breeding facility) rather than on genetics (mouse strain).

We found that, at higher taxonomic levels, the GTM of inbred individuals was dominated by members of the *Bacteroidetes* phylum, whereas *Firmicutes* dominated in the GTM of wild populations. It has been documented that both obesity and efficiency of nutrient harvest are associated with an increasing ratio of *Firmicutes* vs. *Bacteroidetes* in both mice and humans (Ley *et al.* 2005, 2006; Turnbaugh *et al.* 2006, 2009; Jumpertz *et al.* 2011).

Marked difference in the metabolism of wild and inbred individuals is suggested by the fact that KEGG pathways corresponding to lipid and protein metabolism were predicted to be overrepresented in wild vs. inbred individuals. Note, however, that interpretation of differences in metabolic potential between wild and inbred house mice cannot rely solely on the association between obesity and GTM in the laboratory mouse model. For example, we have shown, consistently with Turnbaugh et al. 2006, that several KEGG pathways associated with the obese genotype, such as 'glycerolipid metabolism' or 'ABC transporters', are more abundant in wild populations. On the other hand, other obesity associated pathways, such as 'the Cytrat cycle', 'glycosphingolipid' and 'glycosaminoglycan metabolism' are more abundant in inbred individuals. A number of other pathways known to vary between obese and lean mice, such as 'galactose metabolism', 'pyruvate metabolism', 'starch' and 'sucrose metabolism', did not exhibit significant levels of variation between wild and inbred populations in our study.

A more detailed insight into GTM composition at lower taxonomic scales suggests that relative abundance of ~16% of OTUs differs between wild and inbred mice. These differences are likely to mirror different environmental conditions, such as diet, species pool of environmental bacteria and/or housing. Several Ruminococcus OTUs, for example, enable cellulose digestion in some mammalian species (Michalet-Doreau et al. 2002), and these were more abundant in wild populations, suggesting differences in diet composition. Similarly, three Lactobacilaceae OTUs were more abundant in wild population. Many members of this taxon are important for normal functioning of the digestive tract and immune system as they exhibit anti-inflammatory properties (Mohamadzadeh et al. 2011) and may reduce the severity of bacterial infection in the digestive tract (Naaber

et al. 2004). Depletion of these bacteria in the GTM is associated with 'western lifestyle type diseases' such chronic inflammatory disorders (Michail 2009; Harb et al. 2013) and diabetes (Honda et al. 2012). Interestingly, in line with these findings, diabetes-associated KEGG pathways were predicted to be overrepresented in inbred samples within our study.

Currently, laboratory mouse strains of different origin (both with respect to genetic background and source breeding facility) are used in experimentally focused biomedical research. Both Ivanov *et al.* (2009) and Friswell *et al.* (2010) have previously reported that, despite being isogenic, mouse strains of different provenance possess different GTM (see also Rogers *et al.* 2014), which could contribute to conflicting results in some manipulative experiments (reviewed in Hooper *et al.* 2012). It follows, therefore, that there is a need for the establishment of isogenic mouse strains with standardized GTM (Orcutt *et al.* 1987; Hooper *et al.* 2012).

Our results indicate a pronounced difference in GTM composition and its metabolic potential between wildderived inbred strains and wild house mouse populations. Given the profound effect of GTM composition on host phenotype (Bäckhed et al. 2005; Gaskins et al. 2008; Sekirov et al. 2010; Jumpertz et al. 2011; Koch & Schmid-Hempel 2011; Cryan & Dinan 2012; Hooper et al. 2012), it is probable that at least some variation between wild and inbred individuals is influenced by this factor. We suggest, therefore, that differences in wild and inbred GTM should be taken into account when undertaking biomedical research studies that use house mice as an animal model for studying the association between GTM and various metabolic, neural and behavioural disorders. In particular, further research should be undertaken to validate the usage of individuals with captive bred GTM as appropriate control groups for this type of experiment. Alternatively, based on our results, experimental groups that mirror the composition and interindividual variability found in wild GTM may further improve the robustness of such studies.

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J.P. provided samples; J.P. and J.V. contributed reagents; D.C. and J.K. conceived and designed the experiments; D.C., J.V. and J.K. performed the experiments; J.K. analysed the data; and J.K., D.C. and J.P. wrote the manuscript.

# Data accessibility

Demultiplexed unfiltered ssf files: http://www.ebi.ac. uk/ena/data/view/PRJEB6634 Clusters of quality filtered sequences and OTU table: Dryad doi:10.5061/dryad.018 g4.

# Supporting information

Additional supporting information may be found in the online version of this article.

Table S1 Detailed listing of house mouse samples used

**Table S2** Taxonomic classification of dominant OTUs and their variation between sample groups

**Table S3** Variation of predicted KEGG pathways between sample groups

Appendix S1 Taxonomic composition of house mouse samples

Appendix S2 Extended statistical analysis