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Segregational drift and the interplay between plasmid copy number and evolvability

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Abstract

The ubiquity of plasmids in all prokaryotic phyla and habitats and their ability to transfer

between cells marks them as prominent constituents of prokaryotic genomes. Many plasmids

are found in their host cell in multiple copies. This leads to an increased mutational supply of

plasmid-encoded genes and genetically heterogeneous plasmid genomes. Nonetheless, the

segregation of plasmid copies into daughter cells during cell division is considered to occur in

the absence of selection on the plasmid alleles. Here, we investigate the implications of

random genetic drift of multicopy plasmids during cell division – termed here *segregational*

drift – to plasmid evolution. Performing experimental evolution of low- and high-copy non-

mobile plasmids in *Escherichia coli*, we find that the evolutionary rate of multicopy plasmids

does not reflect the increased mutational supply expected according to their copy number. In

addition, simulated evolution of multicopy plasmid alleles demonstrates that segregational

drift leads to increased loss frequency and extended fixation time of plasmid mutations in

comparison to haploid chromosomes. Furthermore, an examination of the experimentally

evolved hosts reveals a significant impact of the plasmid type on the host chromosome

evolution. Our study demonstrates that segregational drift of multicopy plasmids interferes

with the retention and fixation of novel plasmid variants. Depending on the selection pressure

on newly emerging variants, plasmid genomes may evolve slower than haploid

chromosomes, regardless of their higher mutational supply. We suggest that plasmid copy

number is an important determinant of plasmid evolvability due to the manifestation of

segregational drift.

Introduction

Plasmids are genetic elements that colonize prokaryotic cells where they replicate independently of the chromosome. They are considered as a major driving force in prokaryotic ecology and evolution as they can be transferred between cells, making them potent agents of lateral gene transfer (Lederberg and Tatum 1946; Thomas and Nielsen 2005) and microbial warfare (Czárán et al. 2002; Majeed et al. 2011). Plasmids may contribute to the evolution of their host via the supply of genes encoding for new functions that are beneficial under growth limiting conditions (e.g., antibiotics resistance (Porse et al. 2016), resistance to heavy metals (Gullberg et al. 2014; Dziewit et al. 2015)) or during adaptation to new habitats (e.g., catabolic functions (Schmidt et al. 2011)). Nonetheless, the importance of plasmids goes beyond microbial evolution as they are widely used as vectors for genetic engineering (Simon et al. 1983; Bevan 1984) as well as for applications in biotechnology (Ullrich et al. 1977) and synthetic biology (Shetty et al. 2008). Therefore, understanding plasmid biology and evolution is of major interest for basic microbiology research and biological applications.

Plasmids differ from the chromosomal genetic component in several key aspects, including a relatively small genome size and a variable ploidy level. Plasmids are often found in multiple copies in the cell where the plasmid copy number (PCN) depends on the replicon type, the host genetics, and the environmental conditions (Nordström 2006; Santos-Lopez et al. 2017). The copy number of natural plasmids ranges between 1-15 for low-copy plasmids (Bazaeal and Helinski 1968) and can reach 200 copies for high-copy plasmids (Projan et al. 1987). Plasmids used for biotechnological applications are often modified to have a higher copy number, e.g., 500-700 (Vieira and Messing 1982; Rosano and Ceccarelli 2014), in order to maximize the protein expression level of focal genes. Notably, genes encoded on multicopy plasmids residing within a single host cell may comprise multiple alleles due to independent emergence of mutations in the plasmid copies (Bedhomme et al. 2017; Rodriguez-Beltran et al. 2018). An increased plasmid copy number has the potential to

elevate the probability for the emergence of novel mutations in the gene open reading frame.

Consequently, the mutational supply of plasmid-encoded genes is increased for multicopy

plasmids. Indeed, it has been previously shown that multicopy plasmids have the potential to

enable rapid evolution of plasmid-encoded antibiotic resistance genes under strong selective

conditions for the antibiotic resistance (San Millan et al. 2016).

The inheritance of multicopy plasmids to daughter cells depends on their segregation

mechanism. Some plasmids are equipped with an active partition system. Such systems

comprise plasmid-encoded proteins that function in the translocation of plasmid copies

toward the cell poles prior to cell division, thereby ensuring stable inheritance (e.g., *parABS*

system (Abeles et al. 1985; Baxter and Funnell 2014)). In the absence of a partition system,

plasmid segregation into daughter cells depends on the physical distribution of plasmid

copies in the cell during cell division (Wang 2017). Notably, plasmid segregation into

daughter cells in both routes is considered to occur in the absence of selection on the

plasmid allele pool. Consequently, the dynamics of multicopy plasmid alleles in the

population has a constant component of random genetic drift. In other words, the distribution

of plasmid alleles into daughter cells is independent of the allele impact on the host fitness,

also under selective conditions for a plasmid-encoded trait. Here we term this phenomenon

segregational drift. Previous studies suggested that segregational drift plays a role in the

evolution of eukaryotic organelles – mitochondria and plastids (Birky 2001). Similar to

multicopy plasmids, the mitochondrion organelle is found in multiple copies in the cell (e.g.,

~1,700 in mammalian cells (Robin and Wong 1988)), and in addition, each mitochondrion

organelle may harbour multiple copies of the mitochondrial genome (e.g., up to 10 copies

(Satoh 1991)). Thus, the mutational supply on the mitochondrial genome is larger than the

mitochondrial genome size by several orders of magnitude. The combination of high

mutational supply on the mitochondrion genome and random segregation of the mitochondria

during cell division lead to genetic heterogeneity of the mitochondrial genomes within a

single cell (termed 'mitochondrion heteroplasmy' (Birky 2001)). Notably, many of the

mitochondrial alleles are found in very low frequency, thus the mitochondrial genetic

heterogeneity is observed only with the application of deep-sequencing approaches (e.g., He

et al. 2010). The consequences of segregational drift to the evolution of multicopy plasmids

have been so far overlooked. Population genetics theory postulates that the effect of random

genetic drift on allele dynamics is inversely correlated with the effective population size. Thus

in a larger population, the impact of genetic drift on the dynamics of allele frequency is

reduced (see Lanfear et al. 2014 for review). Thus, the implications of segregational drift to

plasmid genome evolution are expected to be dependent on the plasmid population size

within the cell, i.e., the plasmid copy number. Here we investigate plasmid genome evolution

in light of the segregational drift using experimental evolution and *in silico* simulation of allele

dynamics in the population.

Results

Experimental evolution.

In our study we compare the rate of evolution between low-copy and high-copy plasmid

replicons using an experimental evolution approach. In the experiment, two model plasmids

having low- or high-copy number were evolved in an *E. coli* host under conditions selecting

for the plasmid presence. The low-copy model plasmid (pLC, 5.2 kb; Fig. 1) is derived from

plasmid pBBR1 (Antoine and Locht 1992) that is known to stably replicate in a broad host

range despite lacking an active partitioning mechanism. The plasmid encodes an ampicillin-

resistance gene (*bla*; β-lactamase) and is non-mobile (i.e., it is not transmissible via

conjugation). The average pLC copy number in the wild-type *E. coli* host was 8.0 ± 0.78 (*SD*,

n = 12; using qPCR), which is within the range calculated previously for pBBR1 (Jahn et al.

2016). The accessory part of the high copy number model plasmid (pHC, 4.8 kb; Fig. 1) is

identical to that of plasmid pLC, but the backbone comprises a pUC origin of replication

(Vieira and Messing 1982), that is a derivative of the *ori* of plasmid ColE1 (Hershfield et al.

1974). The copy number of plasmid pHC in the wild-type *E. coli* host was approximately 80-

fold higher than that of plasmid pLC (715.56 ± 27.4 *SD*, *n* = 12; using qPCR). This is within

the typical copy number range of the pUC plasmid backbone (Rosano and Ceccarelli 2014).

In addition to PCN, our experimental evolution setup accounts for two known determinants of

evolutionary dynamics: mutation rate and selection pressure. To vary the mutation rate, the

plasmids were evolved in two host genotypes, *E. coli* K-12 substrain MG1655 (wild-type, wt)

and a hypermutator derivative of the wild-type strain ($\Delta mutS$) in which mutations are

expected to arise faster in comparison to the wild-type (e.g., 33-fold (Arjan et al. 1999)). The

populations were evolved under two different environmental regimes: at 37°C and at 42°C.

Growth of *E. coli* at 42°C is expected to entail an acclimation to the elevated temperature

(Guyot et al. 2014), and it has been shown to involve genomic adaptation (Deatherage et al.

2017). The experiment was conducted with six replicates for each combination of the three

experimental evolution factors – plasmid replicon type, host genotype and growth

temperature. The populations were evolved for 800 generations in continuous culture under

selective conditions for ampicillin-resistance. At the end of the experiment, samples of

ancestral and evolved populations were sequenced using a population sequencing approach.

Notably, the two model plasmids differ in their stability within the host. While the pLC replicon

remained unaltered compared to its pBBR1 ancestor, the high copy number of plasmid pHC

comes at the cost of lower plasmid stability due to deletions of stability factors during

construction of the pUC replicon (e.g., lack of the multimer resolution site *cer* originally

present in plasmid ColE1). We assesed the stability of both plasmids in the *E. coli* wild-type

host using the same selective constraints for the plasmid presence as in the evolution

experiment. No plasmid-free segregants of pLC-wt were observed after overnight growth (*n* =

5), whereas the proportion of pHC-free segregants ranged between 1% and 39% (*n* = 5; see

Methods). Plasmid loss during the evolution experiment has the potential to result in a

proportion of plasmid-free cells in the total population. Such plasmid-free cells have a

collective resistance to ampicillin thanks to the presence of β-lactamase secreted from

plasmid carrying hosts in the environment (Vega and Gore 2014). Therefore, we quantified

the proportion of plasmid-hosts in the evolved populations using replica plating of single

colonies. This showed that pLC-hosts accounted for 89% to 100% of the populations in all

conditions and replicates (Supplementary Table S1). The proportion of pHC-hosts in the

evolved wild-type populations ranged between 1% and 78%, while the range of plasmid

carrying hosts in the evolved hypermutator populations was between 35% and 98%

(Supplementary Table S1). In order to increase the resolution of the existing plasmid allele

pool, we sequenced the genomes of each pHC population using two different sampling

strategies. In addition to the sequencing of total populations (i.e., plasmid host cells and

plasmid-free segregants), the genome of each pHC population was sequenced from pooled

samples of 100 colonies of plasmid carrying hosts (see Methods). In the following, the

resulting genetic variants reported for pHC populations are based on the pooled colonies

data unless stated otherwise.

Plasmid genome evolution.

For the identification of emerging mutations, sequencing reads of the ancestral and evolved

populations were aligned to the reference *E. coli* MG1655 genome and to the two plasmids.

Sequencing coverage varied among the populations and the replicons: the chromosomal

coverage ranged between 104x and 300x, while the average coverage on the plasmids was

between 325x and 28,054x (Supplementary Table S1). The high coverage on the plasmid

genomes increases the detection sensitivity for rare and low-frequency variants. Nonetheless,

in order to compare the number of mutations across replicons and populations, the

sequencing reads of each replicon (i.e., chromosome and plasmids) were subsampled to

match the lowest-coverage dataset unless stated otherwise. Furthermore, variants occurring

in both ancestral and evolved populations were excluded.

A comparative genomic analysis of the subsampled sequencing data revealed two

point mutations on the pLC plasmid (Table 1). Both mutations are located in an intergenic

region and their allele frequency (AF) is lower than 0.10 (Supplementary Table S2). The

paucity of plasmid variants should be evaluated against the background of variants on the

host chromosome, where we observed a total of 1,394 point mutations. We calculated the

expected frequency of point mutations on the two plasmids using the chromosomal

substitution frequency, the plasmid genome size and the plasmid copy number. This shows

that mutations are unlikely to occur on either of the plasmid types evolved in the wild-type host due to the combination of low substitution rate and the plasmid small genome size (Table 1). In contrast, the combination of high genetic polymorphism in the hypermutator host with the high copy number of pHC are likely to result in observable genetic variants in the pHC plasmid within the time span of our experiment (Table 1 and Supplementary Table S1). Nonetheless, the subsampled sequencing data for pHC did not reveal any point mutations occurring on that plasmid. Notably, we detected two single nucleotide insertions in pHC intergenic regions with an allele frequency of AF = 0.3 and AF = 0.35. In order to increase the resolution for the detection of presumably low-frequency point mutations, we applied the same variant calling pipeline to the sequencing read datasets without subsampling (i.e., full sequencing coverage). This analysis revealed one point mutation on plasmid pHC at an intergenic position with a sequencing depth of ~9,600 and a low allele frequency of AF = 0.03 (Supplementary Table S2). In contrast, considering the full sequencing depth for plasmid pLC, which is on average ~10x lower than that of plasmid pHC, we observed two additional intergenic point mutations (Supplementary Table S2). Thus, despite the high sequencing depth, the observed number of point mutations on plasmid pHC does not reach the expected frequency of point mutations, even when the full sequence coverage is included in the analysis. On the other hand, the number of point mutations detected on plasmid pLC is higher than what we expected (Table 1).

The number of point mutations on the pHC plasmid was further validated by the analysis of the total pHC population sequencing data (i.e., samples were prepared without selection for plasmid hosts). Considering the full sequencing coverage, we detected one additional point mutation on the plasmid. The intergenic point mutation was observed with a low allele frequency of AF = 0.04 (Supplementary Table S2) in one of the pHC populations having the hypermutator host genotype. In addition, we compared the number of point mutations observed on the chromosome in the pooled pHC host populations to that of the total population sequencing data (Supplementary Table S1). The comparison shows that the wild-type populations include, on average, two additional point mutations per population in

comparison to the corresponding data of pooled plasmid host colonies. The comparison of

hypermutator populations shows that the total population sequencing data includes, on

average, 14 point mutations per population less than the pooled colonies sequencing data.

Overall, the above comparison between the two sequencing approaches demonstrates that

the number of point mutations observed in the pooled host colonies data is representative

both for the number of variants on the pHC plasmid and the chromosome.

This discrepancy between the observed and expected frequency of point mutations

on the pHC plasmid can be explained under the segregational drift hypothesis as due to the

rapid loss of most plasmid mutations. Our analysis further reveals that the allele frequency of

plasmid variants is considerably lower in comparison to that of chromosomal variants. Of the

total chromosomal point mutations, 30% reached fixation within the population (AF $\geq$ 0.9),

whereas point mutations on plasmids remain at a low frequency (AF $\leq$ 0.09; Supplementary

Fig. S2). The stark difference in the allele frequency distribution between the plasmid and

chromosomal alleles indicates that the increase in plasmids allele frequency is considerably

slower in comparison to chromosomal alleles. This observation is in agreement with our

hypothesis according to which mutations that emerge on multicopy plasmid genomes are

repeatedly 'diluted' in the population due to segregational drift.

Plasmid population dynamics.

What are the theoretical expectations for the impact of segregational drift on the dynamics of

plasmid alleles in comparison to chromosomal alleles? To address this question we

simulated plasmid and chromosome allele dynamics. For this purpose, we applied the

standard haploid Wright-Fisher model to simulate forward-time evolution of a population of

cells that is subject to random genetic drift (Gillespie 2010). In the simulation, the cell

population size remains constant throughout evolution. To simulate plasmid evolution we

added an additional layer of random genetic drift by applying a Wright-Fisher model to the

plasmid inheritance (similarly to mitochondria simulated evolution (Peng and Kimmel 2005)).

Thus, during cell division, plasmid alleles are selected randomly into the daughter cells of the

next generation. In terms of population genetics, the plasmid population in the simulated evolution has a substructure within the population. This implies that plasmid allele dynamics have two hierarchical levels: within the cell and within the total population.

We compared the allele frequency dynamics of neutral variants emerging in a population of $N = 10^5$ cells on a haploid chromosome (termed $Cx10^5$) and two multicopy plasmids: one plasmid has a copy number of 10 plasmids per cell (termed $p10x10^5$), while the second plasmid has 100 copies per cell (termed $p100x10^5$). The plasmid copy number remains constant throughout the simulation, hence, a population of 10^5 cells harbours a total of 10^6 p10 replicons. The total number of p100 replicon in such a population is 10^7. In order to examine the effect of the increased total number of plasmid replicons, we simulated in addition the allele dynamics of a haploid chromosome with two additional population sizes. In the first population, the number of cells equals the total p10 replicons, i.e., $N = 10^6$ (termed $Cx10^6$). In the second population the number of cells matches the total $p100x10^5$ replicons, i.e., $N = 10^7$ (termed $Cx10^7$). In addition, we eliminate the effect of replicon size on the probability of mutation by defining the plasmids and chromosomes to have an equal size of 500 loci. The evolution of all replicons was simulated over 1,000 generations in 100 replicate populations. The simulation results reveal that the number of accumulated mutations on the plasmids is higher in comparison to $Cx10^5$ and lower in comparison to chromosomes with the respective population size (i.e., $p10x10^5$ vs. $Cx10^6$ and $p100x10^5$ vs. $Cx10^7$; Fig. 2A). These results demonstrate that the number of accumulating plasmid mutations is lower than the expected number according to plasmid mutational supply (i.e., total plasmid replicons). This discrepancy indicates that most plasmid mutations are lost rapidly after their emergence. Further comparison of allele frequency distribution among replicons shows that $p10x10^5$ and $p100x10^5$ median AF is smaller in comparison to $Cx10^5$, and in comparison to the chromosome of the respective population size (Fig. 2B). This demonstrates that the increase in allele frequency of plasmid mutations is generally slower in comparison to the mutations emerging on a haploid chromosome.

To compare the dynamics of mutant alleles residing on the plasmid or chromosome

while excluding the effect of mutational supply, we simulated the inheritance of a mutant

allele already present in the population at a specified initial frequency (f). Thus, the allele

dynamics in the population are determined by the combination of the initial mutant allele

frequency, random genetic drift and the mutant allele selection coefficient (s). The simulation

of an initially low-frequency mutant allele (f = 0.0001) that is neutral to the fitness of the cell

(i.e., selection coefficient s = 0) shows that the majority of mutant alleles on plasmids are lost

quickly in comparison to to mutant alleles on chromosomes (Fig. 3A). For example, the loss

of the mutant allele on the plasmid in 20% of the replicates was observed at generation 22

for p10x10^5 and at generation 76 for p100x10^5, but much later in the chromosomal replicons:

generation 128 in Cx10^5, generation 139 in Cx10^6 and generation 1,136 in Cx10^7 (Fig. 3A).

Increasing the initial mutant allele frequency 10-fold (f = 0.001) results in a similar rate of

allele loss between the plasmids and Cx10^5 populations. Nonetheless, loss rates of plasmid

mutant alleles remain higher in comparison to chromosomal mutant alleles in simulations

with the equivalent total number of replicons (Fig. 3A; Supplementary Table S4). Overall, our

results indicate that the loss dynamics of plasmid alleles due to segregational drift depend on

the initial mutant allele frequency in the total population and the distribution of the mutant

allele among plasmid hosts. Notably, the initial allele frequencies used in the simulation may

be encountered upon the acquisition of a new plasmid variant in the population, e.g., by

conjugation or transformation. In the case of *de novo* plasmid mutations the initial allele

frequency of the mutant plasmid is expected to be extremely low. Hence, according to our

simulations, in the absence of a strong benefit for the cell, a mutation on a plasmid will be

quickly lost.

Performing the simulations where the allele has a positive selection advantage (s =

0.01) reveals a higher loss rate in p10x10^5 in comparison to p100x10^5 populations, but the

median fixation time is lower for p10x10^5 than for p100x10^5. Notably, the median fixation time

of plasmid alleles (when not lost) is smaller in comparison to chromosome populations of

comparable total replicons when the initial frequency is low (f = 0.0001) albeit the majority of

plasmid populations lost the allele under these conditions (Fig. 3A; Supplementary Table S4).

The median fixation time of plasmid alleles is similar to the chromosomal one when the initial

frequency is increased (f = 0.001) (Fig. 3A; Supplementary Table S4). With high selection

coefficient (s = 0.1), allele loss is smaller than for s = 0.01 with a similar median fixation time

among p10x10^5 and the chromosome populations. Notably, the median fixation time in

p100x10^5 is substantially higher in comparison to all other replicons under these conditions

(Figure 3A; Supplementary Table S4). The longer fixation time of p100x10^5 alleles under

these simulated conditions are well explained by the effect of segregational drift on the rate

of plasmid allele fixation within the host cell. The high selection coefficient acts in favour of

cells harbouring at least one plasmid mutant, such that those cells rise quickly to fixation

within the population. In comparison, the increase of the plasmid allele in the total population

is slower due to the effect of random genetic drift of plasmid alleles during cell division, which

hinders the increase of plasmid allele frequency in daughter cells (Fig. 3B). In other words,

when the plasmid copy number is high, plasmid alleles within the cell remain polymorphic

(i.e., heterogeneous) over a longer time scale, also when the mutant allele is beneficial.

Notably, a plasmid copy number of 100 is within the range of PCNs observed in natural

isolates (Projan et al. 1987; Zhong et al. 2011), hence, the extended fixation time observed

for p100x10^5 is likely to occur also under natural conditions. Since our simulations do not

include intra-host plasmid recombination, the observed differences in allele dynamics

between plasmids and chromosomes are solely due to inheritance and segregational drift.

Overall, the results of the simulation demonstrate that segregational drift of multicopy

plasmids has a considerable impact on plasmid allele dynamics in the population by

increasing the loss rate and fixation time of plasmid alleles.

Evolution of the host chromosome.

The comparison between the evolved and ancestral populations revealed that most of the

evolved genetic variants are found on the chromosome (Table 1 and Supplementary Table

S3). The majority of the chromosomal variants (90%) in our experiment are observed in the

hypermutator populations where the substitution rate is, on average, 10-fold higher in

comparison to the wild-type populations (Table 1). Genetic variants are observed in 638 protein coding genes and ncRNA genes, of which 173 (27%) genes accumulated genetic variants in more than one population. The distribution of common mutated genes reveals a strong signal of parallel evolution among populations evolved under similar settings (Fig. 4A). Of the parallel mutated genes, 60 genes (35%) are specific to populations belonging to one of the eight experimental factor combinations, indicating that the combination of host genotype, plasmid replicon type and temperature had an impact on the evolution of these genes. Furthermore, a total of 50 mutated genes (29%) are shared among hypermutator populations having the same plasmid type, regardless of the growth temperature (Fig. 4A). Hypermutator-pLC populations share, on average, 19.62 ± 2.09 (SD) mutated genes, whereas hypermutator-pHC populations show a lower level of parallelism with 14.45 ± 3.30 shared genes on average (SD). Furthermore, two small clusters of parallel mutated genes are observed in the wild-type hosts, the first cluster comprises the pLC populations and the second cluster includes the pHC populations evolved at 42°C.

A strong signal of parallel evolution is observed also at the resolution of specific mutations where 144 (15%) variants were detected in more than one of the 48 evolved population. The allele frequency of parallel variants is significantly higher than the allele frequency of population specific variants ($P < 0.001$, using two-sample Kolmogorov-Smirnov test, $n_{parallel}$ = 720, $n_{specific}$ = 841). To quantify the contribution of the three experimental evolution factors to parallel evolution of the host chromosome, we compared the allele frequency of parallel genetic variants among all populations. We observed a significant effect of plasmid type ($P = 1.1 \times 10^{-69}$) and temperature ($P = 3.9 \times 10^{-46}$) on the allele frequency variation (using ANOVA on aligned rank transformed allele frequencies (Wobbrock et al. 2011); detailed statistics in Supplementary Table S5). In the wild-type populations, there is a strong interaction between plasmid type and temperature, whereas in the hypermutator populations, the same interaction is observed along with a main effect of the plasmid type (Fig. 4B). Our results thus reveal that plasmid replicon type is a strong determinant of host evolution and it has an equal or larger effect in comparison to growth temperature.

To further identify variants that are characteristic for specific factor combinations, we

tested for the effect of the factors and their interaction on the allele frequency of each parallel

variant (using ANOVA on aligned rank transformed allele frequencies and FDR). This results

in 37 variants whose allele frequency is significantly different among the factor combinations

(Fig. 4C; Supplementary Table S6). A single variant occurred in all wt-pLC populations

evolved at 37°C, where a nonsense mutation of Glutamine into a stop codon occurred in the

rpoS gene. Hence, that gene is likely non-functionalized in these populations. The *rpoS*

encodes the RNA polymerase σ^S-subunit that is the master regulator of the general stress

response in *E. coli* (e.g., Patten et al. 2004). Loss of RpoS function was previously reported

to occur rapidly in nutrient-limited chemostat populations (Notley-McRobb et al. 2002). A

cluster including two non-synonymous substitutions is observed for wt-pHC populations

evolved at 42°C. The substitutions occurred in a gene encoding for a tRNA modification

enzyme (*miaB*) and a DNA polymerase I gene (*polA*). Overall, we identified 11 different *polA*

mutations in 16 pHC-populations and none in pLC-populations (Supplementary Table S3).

Mutations that are characteristic to hypermutator-pLC populations (Fig. 4C) are found

in genes that belong to four main functional categories, including carbohydrate metabolism,

fatty acid metabolism, biofilm formation and stress response. Example genes where amino

acid replacements have occurred are propionyl-CoA synthetase (*prpE*), a part of the

propionate metabolism pathway (Textor et al. 1997), and a fimbrial-like adhesin protein

(*sfmH*) that is known to promote adhesion in *E. coli* (Korea et al. 2010). Two additional non-

synonymous substitutions in genes encoding for an integrase (*intA*) and a glycosyl

transferase (*wcaA*) are common to all six hypermutator-pLC populations evolved at 42°C (Fig.

4C). The evolution of hypermutator-pHC populations is characterized by a majority of non-

synonymous point mutations. These variants are fixed (i.e., AF > 0.9) in all hypermutator-

pHC populations. The mutated genes are classified into three main categories including

carbohydrate metabolism, stress response and DNA replication. Example mutated genes are

DNA polymerase II (*polB*) and DNA gap repair protein (*recF*). Taken together, the differential

acquisition of mutations in hypermutator populations carrying plasmid pLC or pHC suggests

that the plasmid replicon type had a significant impact on the host chromosome evolution.

Discussion

Previous experimental evolution studies of plasmid-host adaptation noted that evolved

variants are mostly observed in the host chromosome rather than the plasmid genome

(Bouma and Lenski 1988; San Millan et al. 2014; Loftie-Eaton et al. 2017), and this is indeed

the case also for our study. Naturally, the mutational supply of a chromosome of the size of *E.*

coli (4.6 Mb) is higher than the mutational supply of our model plasmid genomes.

Consequently, more mutations are likely to arise on the chromosome than on the plasmid.

Notably, in most studies of plasmid-host coevolution (including ours), the selection pressure

is imposed on the maintenance of the plasmid in the host population rather than a specific

plasmid allele (e.g., as in San Millan et al. 2016). Because plasmids depend on the host

machinery for their replication and persistence, the selection for plasmid maintenance is, in

practice, imposed on the host machinery rather than on the plasmid itself. Moreover, we

have to keep in mind that the potential adaptive responses to the selective pressure for the

plasmid maintenance are numerous in the complex host chromosomes (e.g., San Millan et al.

2014; Harrison et al. 2015), but are very limited in the genetically compact plasmid genome.

This, in combination with the higher mutational supply of the chromosome, implies that most

adaptive variants are expected to occur on the chromosome. This can explain the high

frequency of genetic variants observed on the chromosome in our experiment, as well as

others (e.g., San Millan et al. 2014; Harrison et al. 2015; Loftie-Eaton et al. 2017).

The significant effect of plasmid type on the chromosomal allele frequency observed

in our experiment (Figs. 4A,C) suggests that evolution of the host chromosome depends on

the replicon type of our model plasmids, with the origin of replication being a prominent

characteristic thereof. The evolutionary scenario portrayed here is highly relevant for

plasmid-mediated evolution of antibiotics resistant strains (e.g., Sandegren et al. 2011;

Jechalke et al. 2014; Stoesser et al. 2016) as in such incidents, the antibiotics impose a

selection pressure for the plasmid maintenance in the population. Our findings suggest that

the evolution of plasmid-mediated antibiotics resistance is expected mostly to involve the

host adaptation to the plasmid replicon type rather than evolution of the plasmid encoded

trait.

The results from our experimental evolution of model plasmids demonstrate that the

low-copy replicon has accumulated more point mutations in comparison to the high-copy

replicon (Table 1). This observation is in contrary to our expectation according to the

mutational supply on the plasmid genome. This deviation from the expectation can be

explained by the inverse correlation between population size and the impact of genetic drift

on allele dynamics in the population. The low-copy plasmid constitutes a small replicon

population within the cell. The dynamics of allele frequency due to drift in small populations

are expected to have large fluctuations such that novel alleles are quickly lost or fixed

(Bodmer and Cavalli-Sforza 1976). In contrast, the high copy plasmid constitutes a large

replicon population within the cell where fluctuations in allele frequency over time are

expected to be small (Bodmer and Cavalli-Sforza 1976). The impact of segregational drift on

plasmid alleles is best understood in comparison to the chromosomal alleles. While novel

chromosomal alleles emerging on a haploid chromosome are inherited to all daughter cells in

the population, the inheritance of plasmid alleles depends on the segregation pattern of

plasmid copies. Indeed, our simulations demonstrate that the frequency of neutral mutations

emerging on a multicopy plasmid is higher in comparison to a haploid chromosome (Fig. 2),

which is in agreement with the higher mutational supply of multicopy plasmids (San Millan et

al. 2016). Nevertheless, the median allele frequency of mutations arising on multicopy

plasmids is lower than that of chromosomal mutations in a comparable population size (i.e.,

number of replicons). Hence, the allele frequency of mutations residing on multicopy

plasmids is increasing slower in comparison to chromosomal alleles. Following the

emergence of a novel plasmid allele, plasmid segregation during cell division has three

potential outcomes for the plasmid genotype of the daughter cells: 1) loss of the novel

plasmid allele, 2) the presence of polymorphic plasmid alleles in the cell, or 3) fixation of the

novel plasmid allele in the cell (see illustration in Fig. 5). The expected effect of plasmid copy

number on the allele dynamics is best observed in the simulated evolution of beneficial

alleles (i.e., s > 0): on the low-copy plasmid (p10), the loss of plasmid alleles occurs more

frequently in comparison to the chromosome, but the median fixation time is nearly the same

as that of the chromosome. Indeed, previous studies showed that under strong selective

conditions, beneficial mutations on multicopy plasmids can rise quickly to fixation (San Millan

et al. 2016). Notably, our simulation of beneficial alleles reveals that alleles on the high copy

plasmid (p100) remain polymorphic over a longer time scale and have a higher median

fixation time (Fig. 3). For neutral alleles, segregational drift of multicopy plasmids hinders the

increase of plasmid allele frequency; it leads to frequent and rapid mutant allele loss on the

one hand, and to longer fixation time on the other hand. Altogether, our results demonstrate

that the effect of segregational drift on plasmid allele dynamics counteracts the effect of

higher mutational supply accompanying increased copy number. Consequently, the

evolvability of low-copy plasmids is higher than that of high copy plasmids despite the latter

higher mutational supply.

The impact of segregational drift is not expected to be limited to non-mobile plasmids

as mobile (or conjugative) plasmids are also found in multicopy state (e.g., Figurski and

Helinski 1979). Plasmid copy number may be heterogeneous within the population (e.g.,

Münch et al. 2015), especially in the absence of an active partition mechanism, and

furthermore plasmid copy number can be variable during evolution (e.g., San Millan et al.

2016; Santos-Lopez et al. 2017). The evolution of multicopy plasmids entails time spans of

intra-cellular heterogeneity, which a situation also encountered by mitochondrial genomes

within single cells. Our study thus reveals parallel evolutionary concepts between plasmids

and eukaryotic organelles. In summary, a realistic portrayal and modelling of plasmid

genome evolution requires deeper understanding of segregational drift.

Methods

Bacterial strains, growth conditions and DNA techniques

Escherichia coli K-12 substr. MG1655 (DSM No. 18039, DSMZ) and a mismatch-repair deficient *E. coli* MG1655 derivative strain (Δ*mutS*) were used in the evolution experiment. *Escherichia coli* DH5α (Hanahan 1983) was used during construction of plasmids. All strains were routinely grown at 37°C in LB medium supplemented with 100 mg L^{-1} ampicillin when required. The evolution experiment was performed in M9 medium (M9/amp$_{100}$) prepared as described (Sambrook and Russell 2001), supplemented with 0.4% (w/v) glucose, 0.1% (w/v) casamino acids, 20 mg L^{-1} uracil, 0.5 mg L^{-1} thiamine, 0.005% (v/v) Antifoam 204 (Sigma-Aldrich), and 100 mg L^{-1} ampicillin. Phusion polymerase (Thermo Fisher Scientific) was used for amplification of DNA-fragments for cloning and plasmid constructs were verified by restriction analysis and confirmed by sequencing. Linear DNA fragments and plasmids were introduced into *E. coli* strains by electroporation using a Bio-Rad Gene Pulser device with parameters reported previously (Dower et al. 1988). PCR-primers are listed in Supplementary Table S7.

Construction of the host strains (wild-type and Δmut S)

A marker-free Δ*mutS* deletion strain was constructed using the λ Red/ET Quick & Easy *E. coli* Gene Deletion Kit (GeneBridges) according to the manufacturers protocol. Briefly, the *mutS* gene was replaced with a PCR generated DNA fragment (primer pair mutS_KO_F/R) containing a kanamycin resistance marker gene flanked by FRT sites. One kanamycin resistant clone, in which the replacement of *mutS* with the marker-containing cassette was verified by PCR (primer pair PGK_F and mutS_test_R), was chosen for the removal of the kanamycin resistance marker through expression of the site specific FLP recombinase carried on plasmid 707-FLPe (GeneBridges). One resulting kanamycin-sensitive clone that had lost 707-FLPe was chosen and the deletion of the marker-containing cassette was verified by PCR using primer pair mutS_test_F/R.

In order to potentially compare mutations occurring within non-coding segments of DNA, both

plasmids and host strains were equipped with segments of randomly composed, non-coding

DNA. Such a 600 bp-stretch of non-coding DNA (termed art3) was commercially synthesized

(GeneAart Strings Service, Thermo Fisher Scientific) and inserted into the Tn*7*-specific

attTn7 site of the wild-type and Δ*mutS* strain following the method described by(McKenzie

and Craig 2006). Briefly, the art3 fragment was amplified by using primer pair art3_F/R and

cloned into the *Not*I restriction site present in the mini-Tn*7*-transposon carried on the

transposition vector pGRG25 (GenBank acc. no. DQ460223). The resulting plasmid was

delivered into *E. coli* wild-type and the Δ*mutS* strain and a clone carrying the chromosomal

attTn7::miniTn7(art3) insertion was verified by PCR targeting the Tn*7*-insertion site (primer

pair attTn7_F/R).

Construction of plasmids pLC and pHC

Plasmid pBBR1MCS-5 (GenBank accession no. U25061) was used to construct plasmid pLC.

A 1,944-bp fragment comprising the gentamicin resistance gene and the *lacZ*α gene

fragment was excised with *BstB*I and *Bsa*I. The resulting plasmid backbone fragment (2824

bp) was blunted and ligated to a blunted *Kpn*I-*Sac*I fragment containing *bla* with its promoter

obtained from pBluescript SK(+) into which the *bla* fragment had been inserted into the

*Eco*RV site as a blunted *Cla*I-*Not*I fragment from plasmid pKD4 (GenBank acc. no.

AY048743). Next, 600 bp long stretches of random DNA (designated as art1 and art2) were

inserted into the *Eco*RI site (art1) and into the *Xho*I site (art2) of pBBR1-bla, giving plasmid

pLC (GenBank acc. no. MH238456). The high copy plasmid pHC (GenBank acc. no.

MH238457) was constructed as plasmid pLC, but the *oriV* and the *repA* gene were then

replaced by a pUC origin of replication. This was achieved by amplifying the plasmid except

for the region comprising the *oriV* and the *repA* gene (using primer pair inv_pLC_F/R). The

resulting fragment (2,735 bp) was treated with T4-Polynucleotide Kinase (New England

Biolabs) and ligated to a PCR fragment (916 bp) comprising the pUC-origin of replication

amplified from plasmid pCR4-TOPO (Thermo Fisher Scientific) with primer pair pUC-ori_F/R.

 Experimental design and setup

The evolution experiment was set up using a full factorial design of 2 plasmid types × 2 host

genotypes × 2 growth temperatures. Six biological replicates were used for each of the eight

experimental factor groups, resulting in a total of 48 populations that were evolved for 800

generations. The evolution experiment was conducted in chemostats made from off-the-shelf

materials similar in design as previously described for the cultivation of yeast (Miller et al.

2013). The chemostat system was modified to enable bacterial culturing over longer time

periods. Briefly, air breaks were introduced upstream of culturing vessels to prevent bacterial

growth into medium supply lines and culture vessels were equipped with additional effluent

ports to ensure constant efflux in case of clogging caused by biofilm-formation (additional

information provided in Supplementary Fig. S3). Prior to the evolution experiment, plasmid

pLC and pHC, respectively, were transferred into the wild-type and the Δ*mutS* strain and a

single colony of each type (wt-pLC, wt-pHC, Δ*mutS*-pLC, and Δ*mutS*-pHC) was used to

found a population that was serially passaged (1:1000) daily for about 100 generations in

M9/amp$_{100}$ at 37°C to acclimate the populations to the media conditions. At the onset of the

evolution experiment, 50 µl overnight culture of each of the four populations (termed here

ancestral populations) was used to inoculate 6 replicate chemostat cultures with a working

volume of 20 ml at 37°C and 42°C each. The 48 chemostat cultures were grown to stationary

densities overnight (i.e., without addition of media) and then continuously incubated at a

dilution rate of 0.27 h^{-1}. The bacterial concentration of chemostat cultures grown at steady-

state was stable over time and among replicates (M = 1.92 × 10^{8} cells per ml, SD = 1.11 ×

10^{7} cells per ml, n = 432). Nevertheless, in order to ensure the stable maintenance of flow

rates during the course of the experiment, chemostat units and tubings were replaced every

three weeks (approximately 200 generations). Meanwhile, chemostat populations were

stored at -80°C in 25% (v/v) glycerol. From these frozen stocks, the chemostat cultures were

restarted as follows: 10 ml of the frozen stock culture were pelleted by low-speed

centrifugation (5000×g) for removal of glycerol and the cells were harvested in the final

culturing volume of 20 ml and transferred into chemostat vessels. Each population was

evolved for 90 days at steady-state resulting in approximately 800 generations. On a regular

basis, all chemostat populations were reciprocally tested for cross-contamination using

qPCR. To this end, each culture was tested for the presence of the *mutS* gene and for the

presence of both plasmids using the following primer pairs: q_pBB_rep_F/R complementary

to the *rep* gene on plasmid pLC, q_pUC_ori_F/R complementary to the *ori* on plasmid pHC,

and primer q_mutS_F/R complementary to *mutS* gene in the *E. coli* MG1655 background.

The proportion of plasmid bearing cells in evolved and ancestral populations was assessed

by appropriately diluting samples of chemostat cultures with 1x phosphate-buffered saline,

spreading on M9 agar plates and incubating at 37°C or 42°C for 14 hours. At least 100 single

colonies of each chemostat sample were transferred to M9/amp$_{100}$ and to M9 agar plates.

Colonies that grew only on non-selective media were counted as plasmid-free segregants.

To verify the absence of the plasmids, 1-2 randomly chosen colonies per population were

chosen for testing the presence of the plasmid using qPCR (q_bla_F/R for the plasmids and

q_idnT_F/R for the chromosome). The proportion of plasmid-bearing cells was determined

from the comparison of the number of colonies grown on ampicillin plates to the total number

of transferred colonies grown on non-selective medium.

Plasmid copy number and loss frequency

Plasmid copy numbers were determined using quantitative PCR (qPCR) as described

previously (Skulj et al. 2008), but with primers complementary to *idnT* on the *Escherichia coli*

MG1655 chromosome (q_idnT_F/R) and to the *bla* gene (q_bla_F/R) on plasmid pLC and

pHC. Samples of overnight cultures were heated at 98°C for 10 min and then frozen at -20°C

to disrupt the cells. The qPCR reactions were carried out in a total volume of 10 µl containing

1x iTaq Universal SYBR Green Supermix (Bio-Rad Laboratories), 100 nM of each primer

(final concentration) and 1 µl sample. All qPCR reactions, including positive and non-

template controls, were performed in technical replicates on a CFX Connect Real-Time PCR

Detection System (Bio-Rad Laboratories) using the following cycling conditions: 95°C for 3

min, and 40 cycles of 10 s at 95°C and 1 min at 59°C. Primer specificities and efficiencies

were determined using standard-curve and melt-curve analyses. The ratio between the

number of plasmid specific amplicons and chromosome specific amplicons is defined as the

plasmid copy number (Skulj et al. 2008) and was here calculated considering the

amplification efficiencies of both primer pairs.

To determine the segregational loss of plasmid pLC and pHC, five single colonies of

each strain (i.e., wt-pLC, wt-pHC) were picked from M9/amp$_{100}$ agar plates and grown

overnight in 2 ml M9/amp$_{100}$ with constant shaking at 37°C for approximately 4.4 ± 0.42 (*SD*)

generations. Appropriate dilutions of stationary cell cultures were plated on M9 agar,

incubated overnight at 37°C, and 100 colonies per replicate were streaked onto M9/amp$_{200}$,

M9/amp$_{100}$, and M9 plates. The number of plasmid-free segregants was determined from the

comparison of colonies grown on selective and non-selective plates.

Whole genome sequencing

The ancestral populations of the wild-type and the Δ*mutS* strain carrying the designated

plasmids (wt-pHC, wt-pLC, Δ*mutS*-pHC, and Δ*mutS*-pLC) and the 48 populations evolved for

800 generations were sequenced using Illumina sequencing technology. Population

sequencing was used to enable the detection of rare variants and to asses the frequency of

alleles within and across populations. Total DNA (genomic DNA plus plasmid DNA) was

isolated from archived samples stored at -80°C using the Wizard Genomic DNA Purification

kit (Promega). The preparation of DNA for sequencing of the evolved populations differed

between the two plasmids due to the difference in their stability within the host. For pLC-

populations, 1 ml of cell suspension from thawed samples was used for DNA extraction. The

frequent loss of pHC led to the presence of plasmid-free segregants in the evolved

populations. Therefore, in addition to the sequencing of DNA extracted from total pHC-

populations, DNA was sequenced from pooled colonies of plasmid-carrying cells. The deep

sequencing of plasmid-carrying hosts allows the detection of rare variants from a population

of cells with a potentially heterogeneous plasmid allele pool. DNA from total pHC populations

was isolated from archived samples as described for pLC-populations. The preparation of

DNA from plasmid-carrying cells included a preliminary stage of selection for plasmid-hosts.

Thus, frozen chemostat culture samples were first plated on LB plates supplemented with

ampicillin (100 µg ml^{-1}) and grown overnight at 37°C or 42°C, to select plasmid-carrying cells.

Then, 100 colonies per pHC-population were pooled for extraction of total DNA (i.e., 4x600

colonies in total). Sample libraries for Illumina sequencing were prepared using either the

Nextera or Nextera XT library preparation kit. All samples were quantified on a Qubit

fluorometer (Invitrogen by Life Technologies) and DNA fragment length distribution was

assessed on a TapeStation (Agilent Technologies). Libraries were sequenced on either a

HiSeq 2500 platform with 2x125 bp reads or a NextSeq 500 platform with 2x150 bp reads

(see Supplementary Table S1 for details). Due to low initial coverage, some libraries of the

wt-pLC and wt-pHC populations were re-sequenced on a HiSeq 2500 platform.

Variant detection

Sequencing reads were trimmed to remove both Illumina specific adaptors and low quality

bases using the program Trimmomatic v.0.35 (Bolger et al. 2014) with these parameters:

`ILLUMINACLIP:NexteraPE-PE.fa:2:30:10 CROP:150/125` (NextSeq/HiSeq) `HEADCROP:5`

`LEADING:20 TRAILING:20 SLIDINGWINDOW:4:20 MINLEN:36`. Quality of sequencing reads

was inspected before and after trimming using FastQC v.0.11.5 (Andrews 2010). To confirm

plasmid maps derived from partial Sanger-sequencing after plasmid construction, plasmid

genomes of the four ancestral populations were assembled from trimmed paired end

sequencing reads using plasmidSPAdes (SPAdes v.3.9.0 (Nurk et al. 2013)). To improve

assembly speed and accuracy of the high copy plasmid pHC samples, BBNorm (BBMap tool

suite v.35.82 (https://sourceforge.net/projects/bbmap/)) was used to normalize paired end

read data to an average coverage of 200x. The assemblies resulted in contigs with

overhangs at each end. There was no variation between plasmid assemblies of the same

type. Therefore, only one assembly of plasmid pLC and pHC was used as a reference for

variant detection. Plasmid genome assemblies were annotated using SnapGene software

v.2.4 (GLS Biotech). The chromosomal insertion sequence art3 (described above) was

included in the reference genome sequence of *E. coli* MG1655 (GenBank acc. no.

NC_000913.3). Reads were then mapped to this reference genome and to the assembled

plasmid genomes using BWA-MEM v.0.7.5a-r405 (Li and Durbin 2009). Corresponding

mapping files of libraries that were sequenced twice were merged using PICARD tools

v.2.7.1 (http://broadinstitute.github.io/picard/). Mapping statistics were retrieved using

BAMStats v.1.25 (https://sourceforge.net/projects/bamstats/files/).Then, indexing, local

realignment of sequencing reads, removal of read duplicates and ambiguous aligned reads

were performed using PICARD tools, SAMtools v.0.1.19 (Li et al. 2009) and GATK v.3.6

(McKenna et al. 2010) retaining only paired mapped reads with a minimum mapping quality

of 20. To enable a between-sample and between-replicon comparison of measures deduced

from the sequencing reads (e.g., number of mutations) that is not biased by differences in

sequencing coverage, subsampling of reads was performed. Therefore, mapping files were

first split by contig (i.e., plasmid and chromosome) using Bamtools v.2.4.2 (Barnett et al.

2011). Reads of each chromosomal and plasmid BAM file were then subsampled to the

minimal mean coverage of 104x using SAMtools (seed 1). Subsampled and non-subsampled

datasets were then used to call variants. Short indels and SNPs were called using LoFreq

v.2.1.2 (Wilm et al. 2012) and VarScan v.2.4.3 (Koboldt et al. 2009). Pindel version 0.2.5b9

(Ye et al. 2009) was used with options -v 30 -M 2 to detect indels. The maximum number

of reads used to call variants with LoFreq and to call variants with VarScan by using

SAMtools mpileup was set to 15,000,000 for non-subsampled datasets. Variants that were

supported by alternative reads in one direction only were removed. Furthermore, only

variants that were concordantly identified by two or more of the variant calling tools were

retained using BEDTools' intersect utility v2.17.0 (Quinlan and Hall 2010) with parameters -f

1.0 -r -u. BEDTools' intersect utility was then used with parameters -f 1.0 -r -v to

keep only variants present exclusively in evolved but not in the ancestral samples. A custom

python script using pysam (https://github.com/pysam-developers/pysam) and R version 3.4.3

(R Development Core Team 2017) were then used to retain only variants that met the

following criteria: allele frequency ≥ 0.02, depth of coverage ≥ 10, ratio of the numbers of

forward and reverse reads supporting the variant ≥ 0.25, a minimum medium mapping quality

of 40, a minimum medium base quality of 30 and a minimum median length of 30 nt to the

end of the reads supporting the variant. Indels within an homopolymer stretch of ≥ 8

nucleotides and indels with less than 10 reads supporting the variant were excluded.

Variants whose read support at the corresponding position in ancestor samples was not

sufficient to detect a variant (i.e., < 10x) were excluded. Variant genomic location and the

type of mutation, i.e., intergenic, synonymous, non-synonymous, or nonsense, were

annotated using an in-house Perl script. Plasmid copy numbers were inferred from the de-

duplicated, non-subsampled BAM files as the ratio of plasmid to chromosomal mean

coverage. The number of intergenic and synonymous sites from the annotated reference

genome sequences that were used for the mapping of sequencing reads. Intergenic

positions were extracted from the reference sequences and the total number of intergenic

sites was calculated as the sum of intergenic positions whose coverage was ≥ 10x. Potential

synonymous sites were determined for each position within the reference sequences. The

total number of synonymous sites were then calculated for each mapping file using the Nei-

Gojobori method (Nei and Gojobori 1986) considering only those positions with a minimum

sequencing coverage of 10x.

Simulation

Forward-time simulations based non non-overlapping generations with a constant cell

population size were performed with python3.4 and the module simuPOP v1.1.8.3 (Peng and

Kimmel 2005). Two types of cells were simulated independently. Cells with chromosomes

have L haploid loci (`ploidy=1`). Cells with plasmids have a predefined plasmid copy number

`cp` that stays constant throughout the simulation. Each plasmid has L loci, resulting in a total

number of L×cp loci in each cell. This was implemented based on `cp` customized

chromosomes (`chromTypes=[sim.CUSTOMIZED]*cp`), each with L haploid loci. Alleles were

simulated as binary (`alleleType='binary'`). Inheritance mode for the chromosome

corresponds to a Wright-Fisher model (operator `RandomSelection()`). Inheritance mode for

the plasmid corresponds to a Wright Fisher model with mitochondrial inheritance of

organelles (operator `RandomSelection(ops=[sim.MitochondrialGenoTransmitter()])`).

For the simulations with mutation, initial populations (t=0) were homogenous, i.e., all

loci had state 0. The mutation rate in the simulation was constant with 2×10^{-9} per locus per

generation (operator `SNPmutator(u=2e-9)`). For the simulated evolutionary dynamics of

mutant alleles, we simulated the evolution of one binary locus (`L=1`) having a mutant allele

already present in the initial population at a specified initial frequency (*f*) and no new

mutation is applied during simulated evolution. Initial populations were randomly generated

with a particular mutant frequency f (parameter `freq=[1-f,f]` for the method

`initGenotype`). Note that this assigns the mutant allele randomly to each locus,

independent of the state of the other loci in the same cell for plasmid simulations. For the

simulated evolution with selection, the probability to be inherited to the next generation is

proportional to the fitness value of the cell. For cells with a chromosome, a fitness value of

1+s was assigned to allele state 1 and fitness value 1 was assigned to allele state 0. For

cells with plasmids, a fitness value of 1+s was assigned if at least one plasmid has allele

state 1, otherwise the fitness is 1 (i.e., dominant selection mode). Selection was applied

using a fitness field (`infoFields="fitness"`) and the operator `PySelector`. Simulations

were run for a particular number of generations using the `evolve` method.

Data availability

Sequence reads of evolved and ancestral populations are found in SRA (SRA acc.

SRP######). Plasmid sequences are found in GenBank (pLC: MH######, pHC: MH######).

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Author contributions

J.I, I.M. & T.D. designed the experiment. J.I. established and performed the experimental

evolution study and the comparative genomic analysis. A.K. performed the simulated

evolution. P.R. sequenced the genomes. T.W and N.H took part in the experimental work.

C.W., G.L, A.K. and T.D. took part in the computational analysis. All authors interpreted the

results and drafted the manuscript.

Competing interests

The authors declare no competing financial interests and no conflict of interest.

Tables

Table 1. Number of intergenic and synonymous point mutations detected in the subsampled read data (104x). Due to the inherent difficulties in the robust inference of insertions and deletions (indels; e.g., Spencer et al. 2014), indels are excluded from the counts of observed point mutations. Hence, our estimate of mutation frequency is conservative. A detailed list of the reported statistics for each of the six biological replicates belonging to one of the eight experimental factor groups is shown in Supplementary Table S1. We note that additional low frequency point mutations were detected on the pLC plasmid considering the full sequencing coverage (see variant details in Supplementary Table S2). A detailed list of all observed variants on the chromosome is given in Supplementary Table S3. The variant distribution across the replicons is depicted in Supplementary Fig. S1.

Plasmid type	Host genotype	T (°C)	Chromosome		Plasmid		
			Observed number of point mutations	Substitution per site per generation[1] $\times 10^{-9}$ (*Mean ± SD*)	PCN[2] (*Mean ± SD*)	Expected number of point mutations[3] (*Mean ± SD*)	Observed number of point mutations
pLC	Wild-type	37	17	2.5 ± 0.7	13.0 ± 2.1	$10.2 \times 10^{-2} \pm 2.8 \times 10^{-2}$	0
		42	16	2.4 ± 1.8	4.0 ± 2.2	$3.6 \times 10^{-2} \pm 3.6 \times 10^{-2}$	0
	Hypermutator	37	148	22.4 ± 5.2	6.0 ± 1.6	$4.3 \times 10^{-1} \pm 1.3 \times 10^{-1}$	1
		42	87	13.2 ± 3.3	3.6 ± 1.2	$1.6 \times 10^{-1} \pm 9.1 \times 10^{-2}$	1
pHC	Wild-type	37	6	0.87 ± 0.6	111.3 ± 66.2	$3.7 \times 10^{-1} \pm 3.5 \times 10^{-1}$	0
		42	6	0.87 ± 0.8	99.7 ± 23.2	$2.9 \times 10^{-1} \pm 2.7 \times 10^{-1}$	0
	Hypermutator	37	98	14.3 ± 3.8	64.8 ± 20.2	3.3 ± 1.7	0
		42	147	21.4 ± 11.5	89.4 ± 27.5	6.5 ± 4.6	0

[1] The chromosomal substitution rate is calculated as $R_{Chr} = N_{obs} \times (N_{sites} \times N_{gen})^{-1}$, where N_{obs} is the number of observed point mutations, N_{sites} is the number of intergenic and synonymous sites on the chromosome having a coverage ≥10 and N_{gen} is the number of generations in our experiment.

[2] Plasmid copy number (PCN) was estimated from the coverage of plasmid and chromosome replicons (detailed list in Supplementary Table S1).

[3] The expected number of point mutations on the plasmid is calculated as $N_{exp} = R_{Chr} \times N_{sites} \times PCN \times N_{gen}$, where N_{sites} is the number of intergenic and synonymous sites on the plasmid having a coverage ≥10.

Figures

Figure 1. Genetic maps of plasmid pLC and pHC. The plasmids are of comparable genome

size and differ in the origin of replication. The *mob* gene originally found in pBBR1 was non-

functionalized by truncation during plasmid construction (denoted as Ψ_{mob}).

Figure 2. Simulated evolution of plasmids and chromosomes under neutrality with mutation

(500 loci, mutation rate 2 × 10^{-9} per locus per generation). (**A**) Distribution of the number of

mutations after 1,000 generations in 100 replicates. The color of the boxes corresponds to

the combination of replicon and population size as in the legend. (**B**) Joint cumulative

distribution function (CDF) of the allele frequency of all mutations present in the replicates

after 1,000 generations.

Figure 3. Simulated evolutionary dynamics of mutant alleles on plasmids and chromosomes.

Allele frequency dynamics over 2,000 generations in 100 replicate simulated populations. (**A**)

Columns are the five combinations of simulated replicon type and population size, and rows

are the different combinations of selective coefficient (*s*) and initial allele frequency (*f*).

Generations are on the x-axis, proportion of AF of replicates are stacked along the y-axis,

population allele frequency is colour coded. Note that the starting frequency of plasmid

alleles is implemented as the frequency of mutant alleles in the total plasmid population.

Thus, the initial frequency of cells with a mutant allele in the population is higher in the

plasmid simulations compared to chromosome simulations (see Supplementary Table S4 for

expected number of mutant cells). (**B**) Frequency trajectories for plasmid simulations with

initial allele frequency *f* = 0.001. The frequency of mutant alleles is calculated as the number

of mutant plasmid alleles in the total plasmid population. The frequency of mutant cells is

calculated as the proportion of cells with at least one mutant plasmid allele within the

population. The frequency of fixed mutant cells is the proportion of cells in which all plasmids

loci contain the mutant allele (i.e., the plasmid allele is fixed in the cell). Simulation results

are presented for *s* = 0.1 where the difference between mutant plasmid allele dynamics and

mutant cells in the population is pronounced. The results of the simulation where *s* = 0.01 are

presented for comparison; in this setting, dynamics of mutant plasmid alleles and mutant

cells are similar.

(A)

| p10x10^5 | p100x10^5 | Cx10^5 | Cx10^6 | Cx10^7 |

Proportion of populations

s=0
f=0.0001

s=0
f=0.001

s=0.01
f=0.0001

s=0.01
f=0.001

s=0.1
f=0.0001

s=0.1
f=0.001

Generation

■ AF=0 (Loss)   ■ AF≤0.8   ■ 0.8<AF<0.99   ■ AF≥0.99 (Fixed)

(B)

p10x10^5 p100x10^5 p10x10^5 p100x10^5

Frequency

s=0.01 s=0.01 s=0.1 s=0.1

Generation

■ Mutant alleles   ■ Mutant cells   ■ Fixed mutant cells

Figure 4. Parallel evolution of the chromosome. (**A**) A heatmap representing the number of

shared mutated genes among the 48 evolved populations. The number of genes mutated in

population pair *x*, *y* is calculated as the number of genes in which genetic variants (single

base substitutions or indels) are observed in both populations. Cells along the diagonal

present the number of genes where a mutation was detected in the corresponding population

and at least one additional population. Annotation bars at the top and left right present the

experimental factors (color-coded as shown in (C)). (**B**) An interaction plot of the main factors

in the experiment as calculated from the complete set of allele frequencies of parallel

variants (included in Supplementary Table S3). Dots represent means and error bars

represent the standard error of the mean. ANOVA demonstrated significant effects for

plasmid replicon type and temperature along with significant interactions between the three

factors used in the experiment (statistics in Supplementary Table S5). (**C**) Color-coded matrix

of variants whose allele frequency is significantly different among the main factors or their

combination (using ANOVA on aligned rank transformed allele frequencies and FDR;

statistics in Supplementary Table S6). Intragenic variants (11 synonymous and 22 non-

synonymous) are indicated by the gene symbol and, in case of a single base substitution,

coded as [ancestral amino acid][amino acid position][alternative amino acid]. The two

insertion mutations are indicated by a gene symbol and coded as [nucleotide position in

gene][+ inserted nucleotide]. [*] indicates a non-sense codon. The two intergenic variants are

indicated by [i] coded as [ancestral nucleotide][genomic position][alternative nucleotide].

Figure 5. An illustration of the consequences of segregational drift to plasmid allele dynamics. The illustrated plasmid copy number is four and its segregation into daughter cells is balanced (i.e., the daughter cells inherit an equal number of plasmids). In the depicted scenario, one allele emerges on the chromosome (red X) and another allele emerges on the plasmid (magenta) at t_0. Both alleles are considered neutral (i.e., $s = 0$). Changes in population size (N) and the frequency of hosts (f_{hosts}) over three generations are presented. The chromosomal allele is inherited into all daughter cells, such that it is present in the total population. Random segregation of the plasmid genotypes leads to decrease in the number of hosts. At t_3, only a small minority in the population harbours the new plasmid allele. In the absence of selection for the plasmid allele presence, the new plasmid allele is deemed to remain at a very low frequency within the population (or be lost). Note: this is a simplistic example that does not include loss of host cells due to drift and it does not include consideration of recombination among plasmids.

Emergence of a new plasmid allele $\rightarrow$ t_0, N=1, f_{hosts}=1

Plasmid allele dynamics depend on plasmid segregation $\rightarrow$ t_1, N=2, f_{hosts}=0.5

Loss of plasmid alleles due to segregational drift $\rightarrow$ t_2, N=4, f_{hosts}=0.5

t_3, N=8, f_{hosts}=0.375

Fixation of the new plasmid allele in a cell